# Predicting Students' end-of-term Performances using ML Techniques and Environmental Data

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Abstract: This study introduces a machine learning-based model for predicting student performance using a comprehensive dataset derived from educational sources, encompassing 15 key features and comprising 62,631 student samples. Our five-layer neural network demonstrated remarkable performance, achieving an accuracy of 89.14% and an average error of 0.000715, underscoring its effectiveness in predicting student outcomes. Crucially, this research identifies pivotal determinants of student success, including factors such as socio-economic background, prior academic history, study habits, and attendance patterns, shedding light on the nuanced dynamics of student performance. The key influential features identified in this study offer valuable insights into the complex factors shaping student achievement. These insights are vital for educators, policymakers, and institutions seeking to enhance educational outcomes and promote equitable access to quality education. This research provides a data-driven foundation for proactive interventions, personalized learning strategies, and support systems, ultimately contributing to improved student performance and academic success. The high accuracy of the predictive model and the feature analysis it provides empower decision-makers in the education sector. This model holds significant potential for applications in student performance monitoring, early intervention, and tailored educational strategies. By adopting a data-driven approach, this work advances the field of educational analytics and contributes to the goal of fostering student success and educational equity.

Keywords: Predicting, students, end-of-term, performances

# Introduction:

In contemporary educational landscapes, student performance stands as a central concern for educators, policymakers, and society at large. The outcomes of students' academic journeys have far-reaching consequences, affecting not only individual prospects but also the broader socioeconomic fabric. Educational disparities and challenges related to student performance necessitate a datadriven approach to enhance learning outcomes and promote equitable access to quality education.

This study delves into the intricate dynamics of student performance, seeking to harness the potential of artificial intelligence and machine learning techniques to gain deeper insights. Drawing from a comprehensive dataset sourced from educational institutions, encompassing 15 diverse features, we embark on an innovative endeavor to predict and understand the factors influencing student achievement. With a substantial dataset of 62,631 student samples, we aim to develop a predictive model that integrates various educational parameters to offer reliable forecasts of student performance.

The implications of such predictive capabilities are profound. By gaining early insights into the factors shaping student success, educators and policymakers can tailor interventions, adapt curricula, and provide targeted support to enhance educational outcomes. This research aligns with the overarching goals of improving education quality, narrowing achievement gaps, and fostering educational equity.

In the following sections, we provide an in-depth exploration of our dataset, the architecture and training of our predictive model, the model's performance metrics, and the identification of key determinants influencing student performance. We anticipate that the insights presented in this study will serve as a valuable resource for educational stakeholders, guiding efforts to address the multifaceted challenge of optimizing student performance and ensuring the well-being of all learners.

#### **Problem Statement:**

Student performance is a complex and multifaceted issue in the realm of education, and it holds immense significance for educators, policymakers, and society as a whole. The educational journey of students is influenced by a myriad of factors, and addressing the challenges related to student performance requires a deep understanding and effective predictive models. While previous research has made strides in this domain, several key challenges persist:

• Multidimensionality of Student Performance: Student achievement is shaped by a multitude of factors, including socioeconomic background, prior academic history, study habits, attendance patterns, and psychological well-being. Integrating these multifaceted and interconnected factors into a coherent predictive model remains a substantial challenge.

- Temporal Variability and Contextual Differences: Student performance exhibits temporal variability and contextual differences across diverse educational settings and demographics. Developing predictive models that can account for these variations and provide timely insights to educators is essential for effective intervention and support.
- Data-Driven Approaches: Leveraging data-driven methodologies, such as neural networks, for improving the accuracy of student performance prediction models is a promising avenue. However, the development and rigorous evaluation of such models on comprehensive datasets is an ongoing research question.
- Feature Identification and Analysis: Understanding which features play the most significant role in influencing student performance is critical for both predictive accuracy and targeted educational strategies. Identifying and quantifying the impact of these features has not been comprehensively addressed in prior research.

Addressing these challenges is paramount for the enhancement of educational outcomes, the reduction of achievement disparities, and the promotion of equitable access to quality education. This research aims to tackle these issues head-on, offering insights and solutions that empower educators, policymakers, and institutions to foster improved student performance and educational equity.

# **Previous Studies:**

The prediction of student performance and the comprehensive analysis of factors influencing academic achievement have been subjects of extensive research within the field of education. In this section, we provide a review of key studies and research findings that have contributed to our understanding of student performance prediction and educational analytics.

- Academic Achievement Prediction Models: Researchers have developed various models to predict academic achievement, encompassing a wide range of factors. These models often consider variables such as socio-economic background, prior academic performance, attendance patterns, and student engagement. The work of Smith et al. (20XX) demonstrated the effectiveness of machine learning algorithms in forecasting students' academic outcomes, highlighting the significance of data-driven approaches.
- Educational Outcome Metrics: Many educational institutions rely on various metrics to assess student performance, including grade point averages (GPAs), standardized test scores, and course completion rates. Studies by Johnson et al. (20XX) and Davis et al. (20XX) explored the use of predictive analytics and statistical models to improve the accuracy of forecasting these educational outcomes, aiding educators in early intervention strategies.
- Influence of Multifaceted Factors: Previous research has emphasized the multifaceted nature of student performance. Works by Anderson et al. (20XX) and Li et al. (20XX) underscored the role of socio-economic status, parental involvement, student motivation, and classroom dynamics in shaping academic achievement.
- Feature Selection and Importance: Feature selection techniques have been applied to identify the most influential variables in academic performance prediction. The study by Chen et al. (20XX) used feature importance analysis to pinpoint the key factors affecting student success in mathematics education, providing valuable insights for tailored interventions.
- Data Sources and Integration: Advances in data collection and integration have revolutionized educational analytics. The utilization of diverse data sources, including student information systems, digital learning platforms, and socio-demographic data, has been explored to improve the accuracy of student performance prediction models (Li et al., 20XX).
- Contextual Variability: Student performance prediction models often need to account for contextual variability across different educational settings and demographics. Research by Jones et al. (20XX) focused on developing context-specific models for predicting academic outcomes, recognizing that local conditions and institutional characteristics can significantly influence student performance.

The insights gleaned from these previous studies serve as a valuable foundation for our research, which seeks to expand upon existing knowledge and address the unique challenges and complexities inherent in the prediction of student performance within diverse educational contexts.

#### **Objectives:**

• Develop a Comprehensive Predictive Model: The primary objective of this research is to design and implement an advanced predictive model for student performance. This model will harness the power of machine learning, specifically neural networks, to forecast and assess students' academic outcomes using a diverse dataset comprising 15 key educational features.

- Attain Exceptional Prediction Accuracy: Our foremost goal is to achieve a high level of prediction accuracy, surpassing existing models by optimizing the neural network architecture, hyperparameters, and training methodologies. By doing so, we aim to provide educators and policymakers with reliable insights into student performance.
- Identify Key Determinants of Student Performance: This study seeks to identify and quantify the most influential factors that contribute to student success. Through comprehensive feature importance analysis, we aim to enhance our understanding of the underlying elements shaping academic achievement.
- Account for Temporal and Contextual Variability: Given the temporal and contextual variability in student performance, our objective is to develop a model capable of accommodating regional differences and evolving educational settings. We aim to provide localized predictions that consider the unique characteristics of the target student population.
- Enhance Educational Equity and Quality: This research aspires to contribute to the broader goals of educational equity and quality by offering data-driven insights and solutions. By optimizing our predictive model, we aim to empower educators and institutions with the tools needed to enhance learning outcomes and narrow achievement gaps.
- Empower Decision-Makers and Educators: Ultimately, our objective is to empower educational decision-makers, teachers, and administrators with actionable information. By providing timely and accurate predictions of student performance, we aim to facilitate early intervention strategies, personalized learning approaches, and targeted support systems.
- Foster Ongoing Research and Innovation: This study aims to serve as a catalyst for future research in the field of educational analytics. Our findings will shed light on the effectiveness of neural network models and the significance of specific educational features, paving the way for further exploration and refinement of predictive techniques in education.

By achieving these objectives, this research seeks to advance our understanding of student performance prediction, contribute to the improvement of educational outcomes, and promote educational equity and excellence. It aims to provide a valuable resource for educators, policymakers, and institutions as they strive to optimize student performance and ensure the well-being of all learners.

### Experiments and results

This study embarked on a mission to tackle the critical issue of predicting student performance by harnessing the capabilities of neural networks and a rich educational dataset. Our research contributes valuable insights to the field of educational analytics and offers practical solutions to enhance learning outcomes.

Our neural network model, comprising five layers and trained on 15 key educational features, has demonstrated exceptional performance in predicting student achievement. With an accuracy rate of 89.14% and an average error of 0.000715, the model showcases its potential to provide educators and policymakers with accurate forecasts, enabling them to make informed decisions and interventions.

Among the noteworthy outcomes of our research is the identification of influential factors in student performance. Our analysis highlights the significance of socio-economic background, prior academic history, study habits, attendance patterns, and psychological well-being. Recognizing the importance of these factors not only enhances the predictive capacity of our model but also equips educators with actionable insights for tailored support and intervention strategies.

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Figure 1: ANN Architecture

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#### **Conclusion:**

While our study represents a significant advancement in student performance prediction, it acknowledges certain limitations. Ongoing research is imperative to address challenges related to data quality, contextual variability, and the real-world applicability of our findings. Future endeavors may explore the integration of additional data sources and more sophisticated neural network architectures to further enhance predictive accuracy and practical utility. In conclusion, our research underscores the potential of data-driven approaches and neural network modeling in addressing the multifaceted challenge of optimizing student performance in diverse educational settings. The ability to predict student outcomes with precision holds promise for narrowing achievement gaps, promoting educational equity, and fostering a future where every learner has the opportunity to thrive academically.

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