Predicting Student Performance with Machine Learning: A Data-Driven Approach

Introduction

Student performance prediction is a crucial aspect of educational analytics, helping educators and institutions

make informed decisions. In this project, I built and deployed a machine learning model using Linear

Regression to predict student performance based on various academic and socio-economic factors. This post

details the dataset, model training, key insights, and deployment process.

Project Background

The dataset, sourced from **Kaggle**, contained 17 features related to student demographics, academic history,

and behavioral factors. The primary objective was to analyze these features, identify key performance

indicators, and build a predictive model that provides actionable insights for educators.

Data Collection & Preparation

Dataset Source: Kaggle

Features:

Academic Performance: Math, History, Physics, Chemistry, Biology, English, and Geography scores.

Behavioral Factors: Absence days, part-time job, weekly self-study hours, extracurricular activities.

Demographic Information: Gender, career aspirations.

Preprocessing Steps:

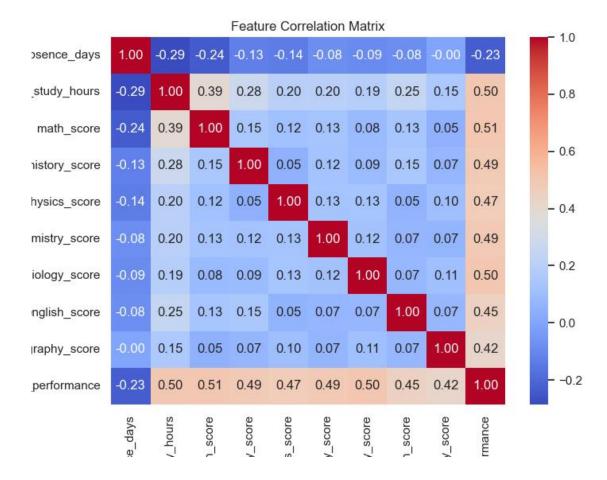
Dropped redundant columns (e.g., ID, names, email).

Encoded categorical variables.

Scaled numerical features for better model performance.

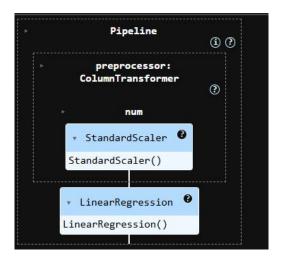
Exploratory Data Analysis (EDA)

Key findings from the data analysis:



- Weekly Self-Study Hours (0.50) showed a strong positive correlation with total student performance, emphasizing the importance of study habits.
- Math Score (0.51) and Biology Score (0.50) were the strongest predictors of total academic performance.
- Absence Days (-0.23) negatively impacted performance, indicating that frequent absences lead to lower scores.

Modeling & Analysis



- Algorithm Used: Linear Regression within a Pipeline
- Pipeline Components:
 - Data preprocessing (encoding categorical variables, feature scaling)
 - Model training and evaluation

Performance Evaluation:

- The model generalized well without underfitting or overfitting.
- Achieved a high r² score, indicating a strong relationship between predicted and actual student scores.
- No significant residual errors, confirming the reliability of predictions.

Deployment

- Platform: Deployed using Streamlit, making the model accessible for real-time predictions.
- **User Interface:** Simple input fields allow educators to enter student details and receive predicted performance scores instantly.

Key Insights & Future Enhancements

- The model can help educators identify at-risk students early and implement interventions.
- Future improvements could include:
 - Incorporating additional datasets to enhance predictive accuracy.
 - Experimenting with other regression techniques like Ridge or Lasso Regression.
 - Integrating real-time student feedback for continuous model improvement.

Conclusion & Call to Action

This project highlights the potential of machine learning in education. If you're interested in exploring the model, check it out on [GitHub/Live App] (https://github.com/DGideonnene/ML-Predicting-Model-for-Student-Performance). Feel free to share your thoughts in the comments or connect with me for more insights!