

Student Dropout and Academic Success Prediction

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Overview

This project aims to analyze and predict undergraduate students' academic outcomes, focusing on dropout, continued enrollment, and graduation. Using machine learning and statistical techniques, the project identifies key factors influencing student success and provides actionable insights to reduce dropout rates.

Goals

- 1. Identify the primary factors influencing student academic outcomes.
- 2. Build predictive models for classifying student statuses into "Dropout," "Enrolled," and "Graduate."
- **3**. Validate the relevance of academic, demographic, and socio-economic features through statistical methods and visualizations.

Dataset Description

The **Student Dropout Prediction Dataset** contains **36 features** describing various academic, demographic, and socio-economic characteristics of undergraduate students, with a total of **4.424 rows**.

Features:

- I. Academic Attributes:
 - **Grades:** Admission grades, semester grades, and curricular unit evaluations (1st and 2nd semester).
 - **Credits:** Curricular units credited, enrolled, and approved.
 - **Performance Metrics:** Academic success indicators like grades and approvals.
- II. Demographic Attributes:
 - **Personal Information:** Age, gender, nationality, marital status.
 - **Parental Information:** Parents' qualifications and occupations.
- III. Socio-Economic Attributes:
 - **Financial:** Tuition fee payment status, scholarship holding status.
 - **External Factors:** Unemployment rate, inflation rate, and GDP.

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Data Preprocessing

I. Handling Missing Values

No missing values detected in the dataset.

II. Feature Engineering:

Categorical variables were encoded, and numerical variables were normalized where necessary.

III. Outlier Removal:

Outliers identified using Z-scores and visualized through boxplots.

IV. Class Balancing:

Applied SMOTE (Synthetic Minority Oversampling Technique) to balance the dataset with 2,209 samples for each class.

Exploratory Data Analysis (EDA)

I. Insights from Visualizations:

- Over 90% of students pay tuition fees on time.
- Female students comprise more than 60% of the dataset.
- Most students enrolled between ages 18-20.
- Boxplots showed differences in semester grades among student categories.
- Heatmaps revealed strong correlations among semester grades and approved units.

II. Key Findings:

- Students with higher admission grades (above 140) have higher graduation rates.
- Semester grades significantly influence academic success.

Feature Selection and Dimensionality Reduction

SelectKBest Result:

Top predictors include semester grades, tuition fee status, and scholarship holding status.

PCA Results: Top 10 principal components explained over 99.9% of variance, ensuring efficient dimensionality reduction.

Model Building and Evaluation

Models Used:

- Decision Tree
- Random Forest
- Logistic Regression
- Support Vector Machine (SVM)

Best Performing Model:

Random Forest Classifier

Accuracy: 83%F1 Score: 84%

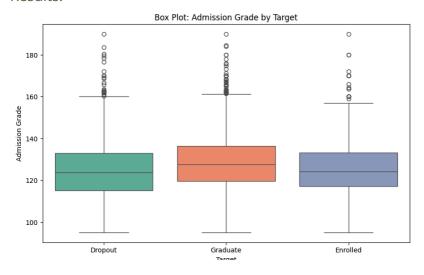
Cross-Validation:

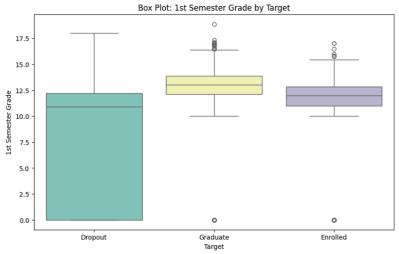
Ensured model reliability with a mean accuracy of 82.5%.

Confusion Matrix:

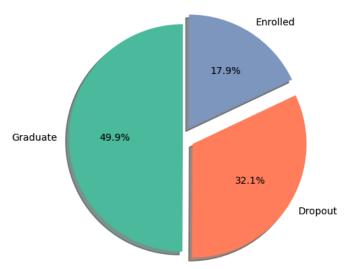
Visualized classification performance with balanced accuracy across classes.

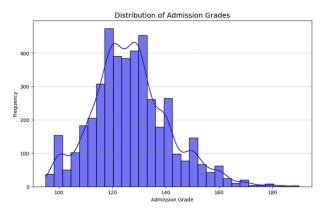
Results:

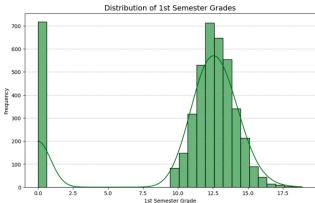


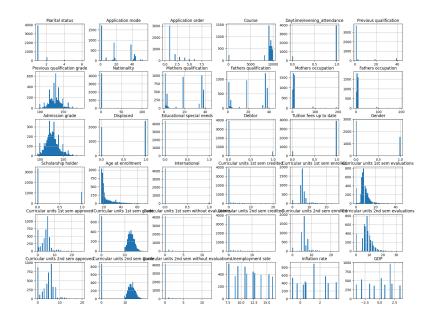


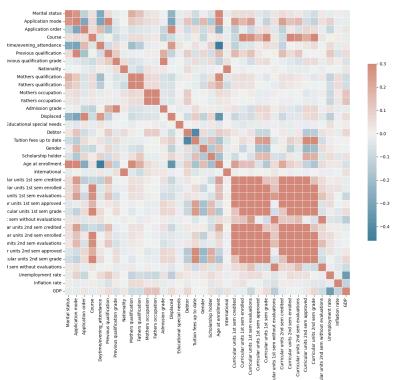
Proportion of Target variable

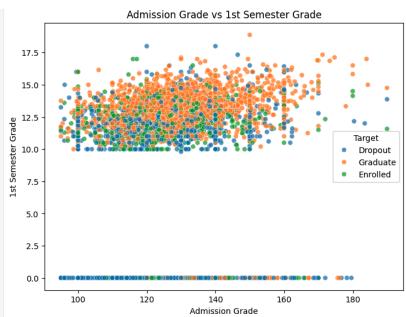


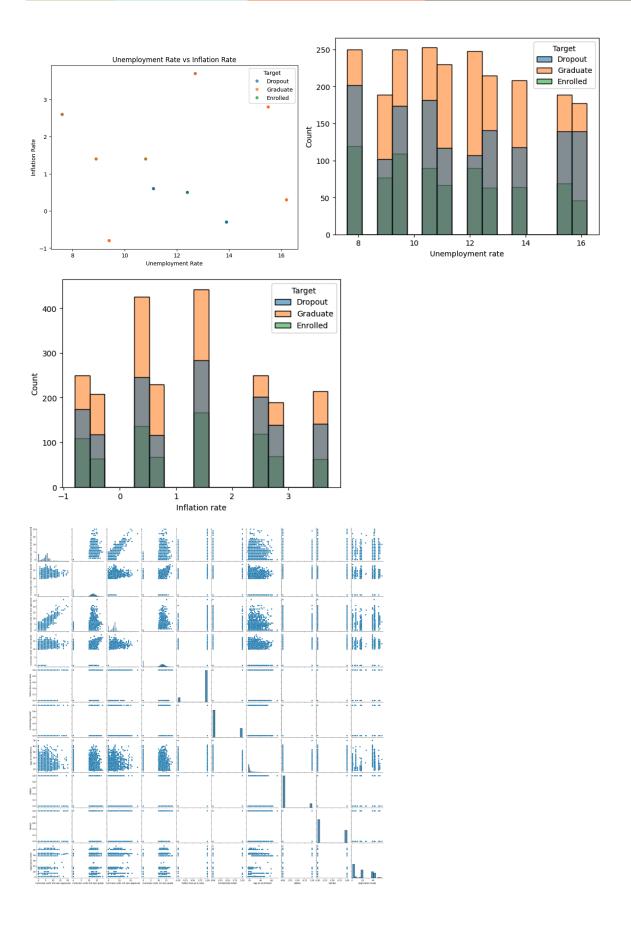


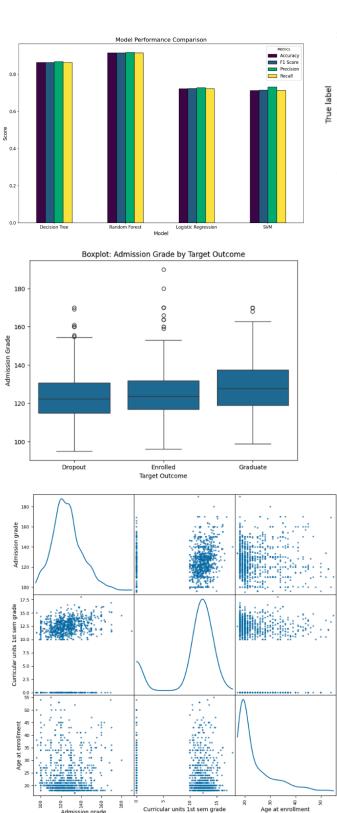


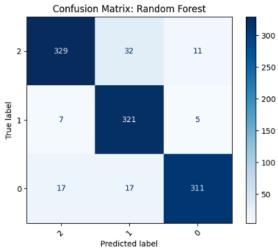


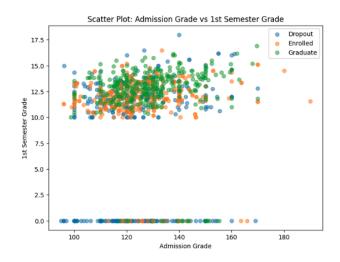


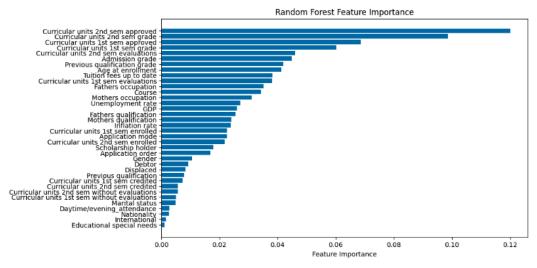












Conclusion:

The analysis identified key factors influencing student outcomes, with "Curricular units 2nd sem approved" (Score: 2110.25) and "1st sem grade" (Score: 973.41) being the most significant predictors. Random Forest achieved the highest performance, with an accuracy of 83% and F1 Score of 84%, effectively classifying student statuses. SMOTE oversampling balanced the dataset (2209 samples per class), improving model reliability. Visualizations highlighted that students with admission grades above 140 were more likely to graduate, while those below 120 had higher dropout rates. These findings emphasize the importance of academic and financial support to enhance graduation rates.

Video URL link:

https://drive.google.com/file/d/1d5nn4OAMdlBxWsfsbZdBDrbXGoYRBg1Y/view?usp=sharing