

Exploring And Predicting CGPA of Bangladeshi Students

Team 2:

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Research Question

General Question:

Can we predict student CGPA based on learning behavior, demographics, and academic history?

Research questions:

- RQ1: Can CGPA be predicted from academic history?
- RQ2: Can CGPA be predicted from demographic background?
- RQ3: Can CGPA be predicted from learning-behavior patterns?



Previous Research Insights

University Records

The strongest predictors of academic performance

“Students' mid-term-exam grades are an important predictor to be used in predicting their final-exam grades” (Yağcı 2022)

Demographics

Add only marginal value when stronger features are present

“Only admission marks and final marks are used; no socio-economic or demographic features are considered.” (Asif et al. 2017)

Learning Behavior

Often depending on verified data and features metrics that enable accurate grade prediction

“the amount of study only emerged as a significant predictor of cumulative GPA when the quality of study and previously attained performance were taken into consideration” (Plant et al. 2005)

Why the research important?

This research helps **validate insights** and **challenge** existing assumptions about **predicting grades** based solely on **self-reported learning habits**.



Data Overview

1,195

Students

Computer Science and
Engineering students in
Bangladesh

31

Features

Self-reported university records,
study habits, technology access
and background traits

2021

First Collection Year

Contains information about
students who began studies
between 2013-2023

Research Main Features



Academic History

Previous academic performance records



Learning Behavior

Study habits and learning patterns



Demographic

Background and personal characteristics

Methodology

Data Cleaning



Feature engineering



Analyze features and
selection



Conclusion and future
work



Compare to previous
research insights



Fit Model and analyze

Methods And Evaluation

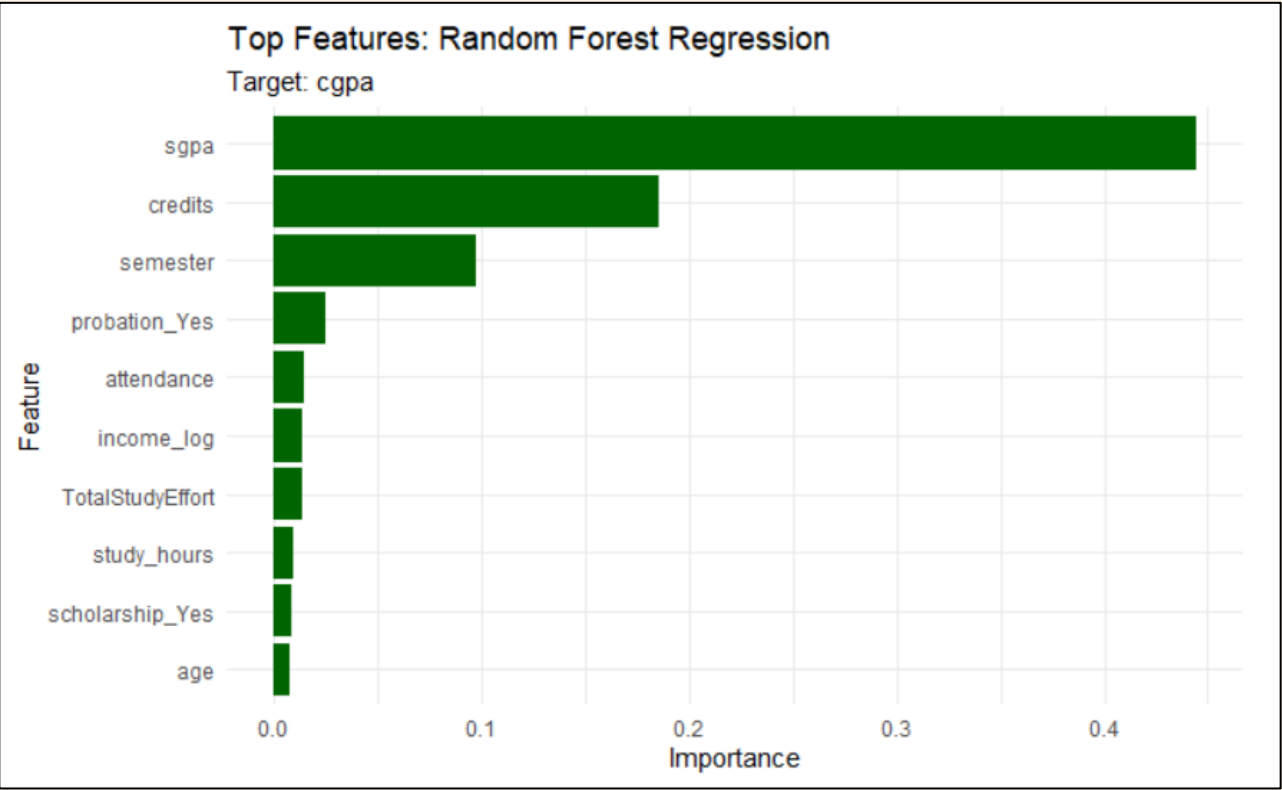
| | Linear Reg. | RF Reg. | RF Classifier |
|--------------------|------------------------------|------------------------------|------------------------------------|
| Valuate Indicators | R^2 , R^2 Adj, RMSE, MAE | R^2 , R^2 Adj, RMSE, MAE | $F1$, Accuracy, precision, Recall |
| Target feature | CGPA (0-4) | CGPA (0-4) | CGPA_CLASS (A-D) |

Results

General Question Results:

| Model Type | Task Type | Metrics | Feature Importance |
|---------------|----------------|--------------------------------|---------------------------|
| Random Forest | Regression | $R^2 = 0.8706$, RMSE = 7.4898 | sgpa, credit, semester |
| Random Forest | Classification | F1 = 0.7365, Accuracy = 0.7368 | sgpa, credits, income_log |

Feature Importance:



Results (Continued)

RQ1 Results:

| Model Type | Task Type | Metrics | Feature Importance |
|---------------|----------------|--------------------------------|---------------------------------|
| Random Forest | Regression | $R^2 = 0.9245$, RMSE = 0.2163 | <u>sgpa</u> , credits, semester |
| Random Forest | Classification | F1 = 0.7822, Accuracy = 0.7812 | <u>sgpa</u> , credits, semester |

RQ2 Results:

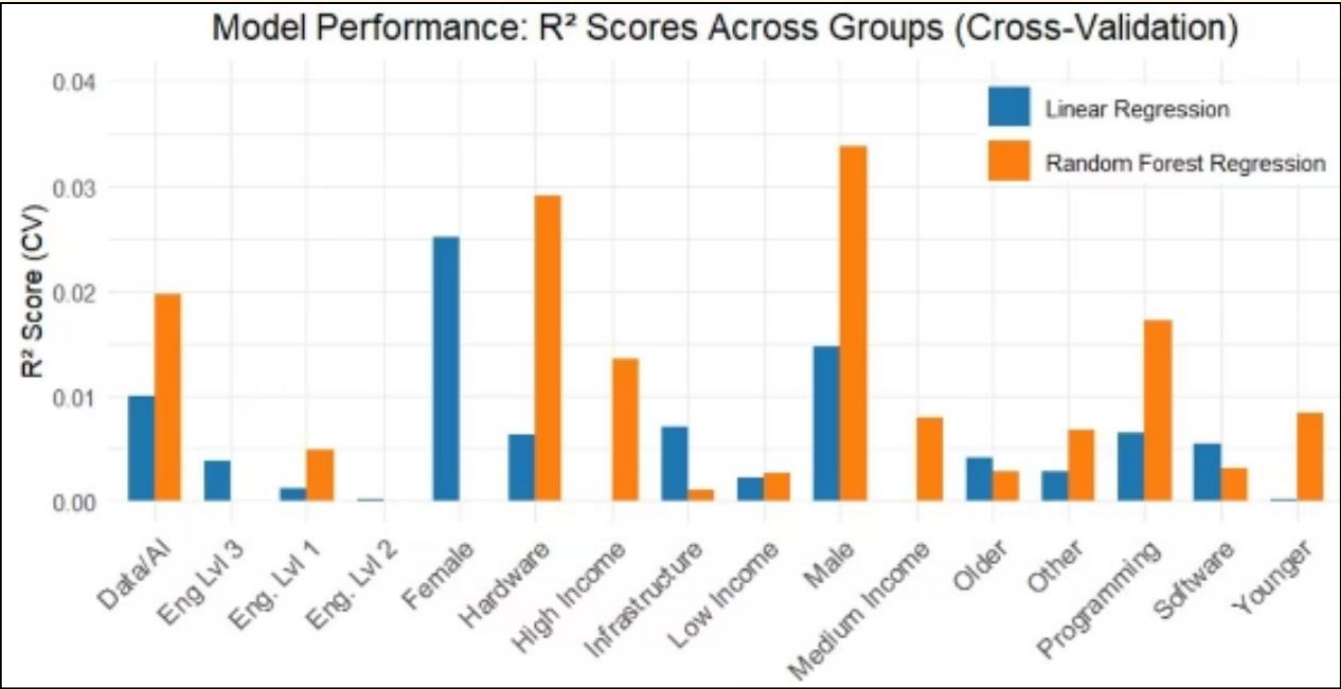
| Model Type | Task Type | Metrics | Feature Importance |
|-------------------|----------------|---------------------------------|---|
| Linear Regression | Regression | $R^2 = 0.0418$, RMSE = 19.2894 | <u>english_Basic</u> , <u>english_Intermediate</u> , relationship_Married |
| Random Forest | Classification | F1 = 0.3297, Accuracy = 0.3333 | relationship_Married, <u>english_Basic</u> , <u>english_Intermediate</u> |

Results (Continued)

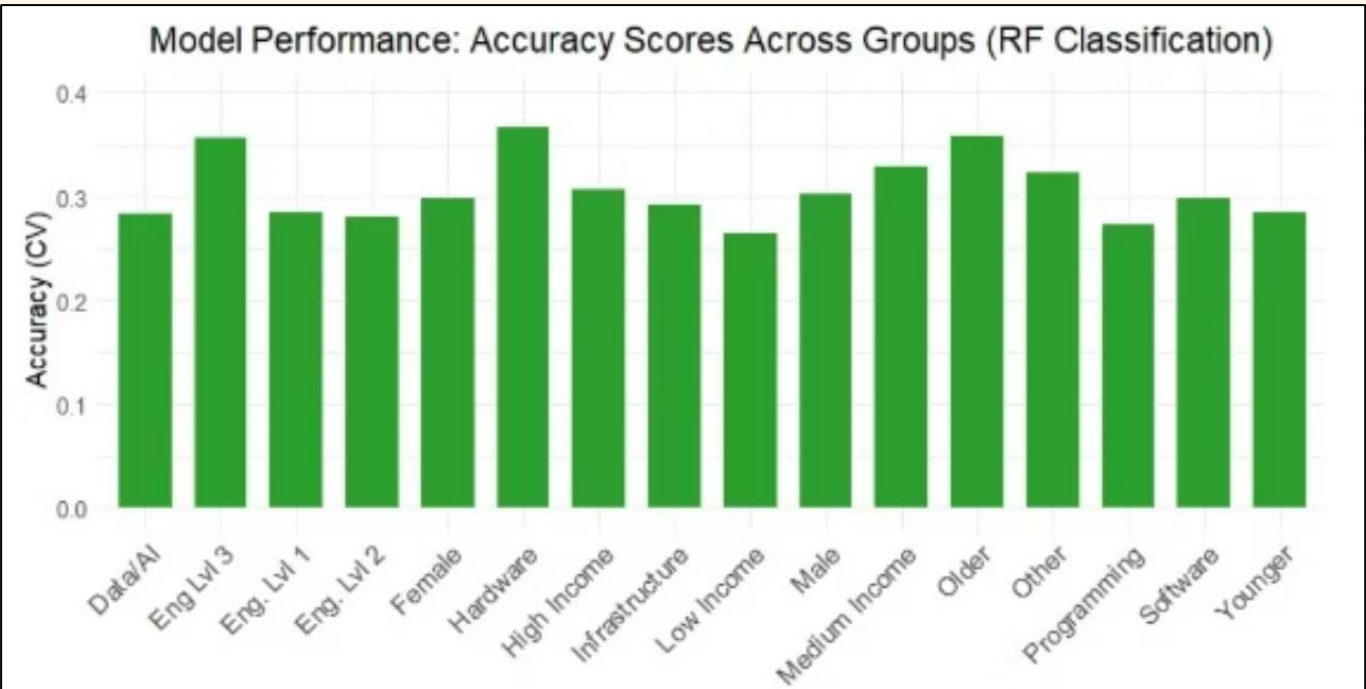
RQ3 Results:

| Model Type | Task Type | Metrics | Feature Importance |
|-------------------|----------------|---------------------------------|-------------------------------------|
| Linear Regression | Regression | $R^2 = 0.0008$, RMSE = 19.2894 | attendance, study_hours, study_freq |
| Random Forest | Classification | F1 = 0.3297, Accuracy = 0.3333 | attendance, study_hours, study_freq |

Sub-group Accuracy (CV)



Sub-group R^2 (CV)



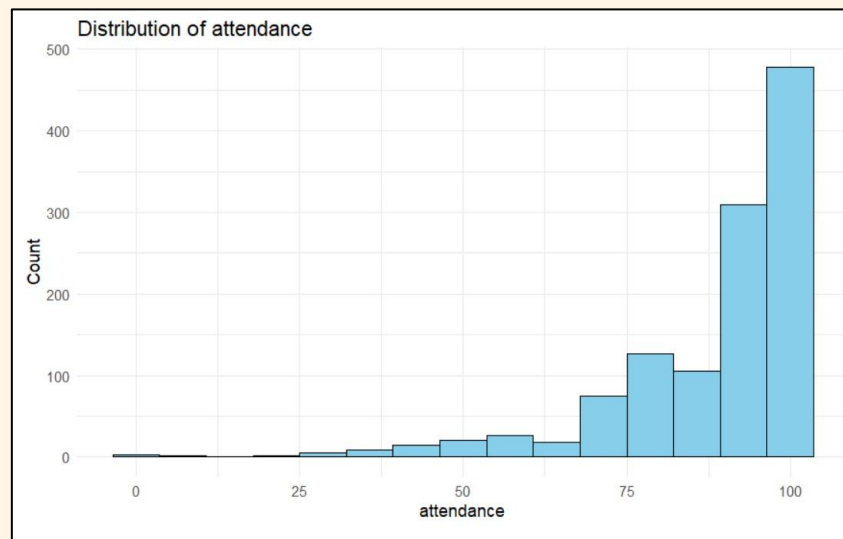
Limitations

- 1 Self-report bias
- 2 Small variance of variables (narrow range)
- 3 Unverified data that doesn't necessarily indicate learning quality

65%

Attendance

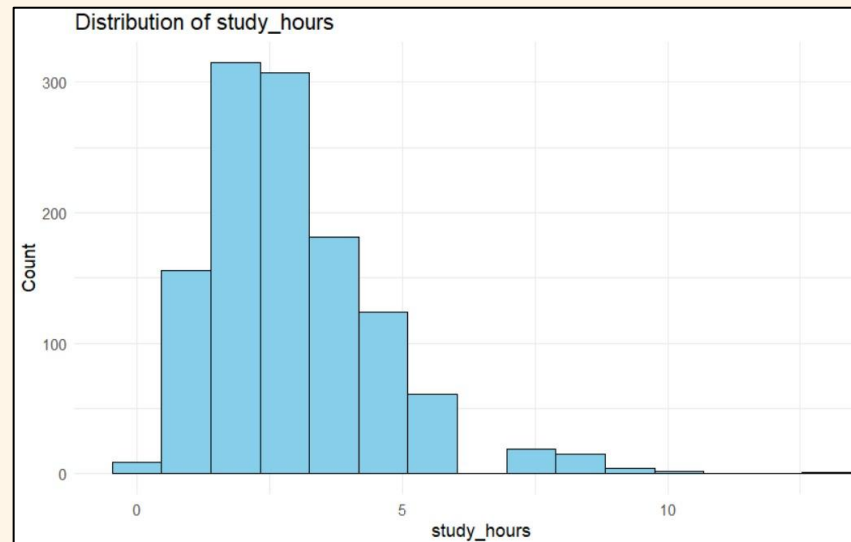
Students report attendance $\geq 90\%$



80%

Study Time

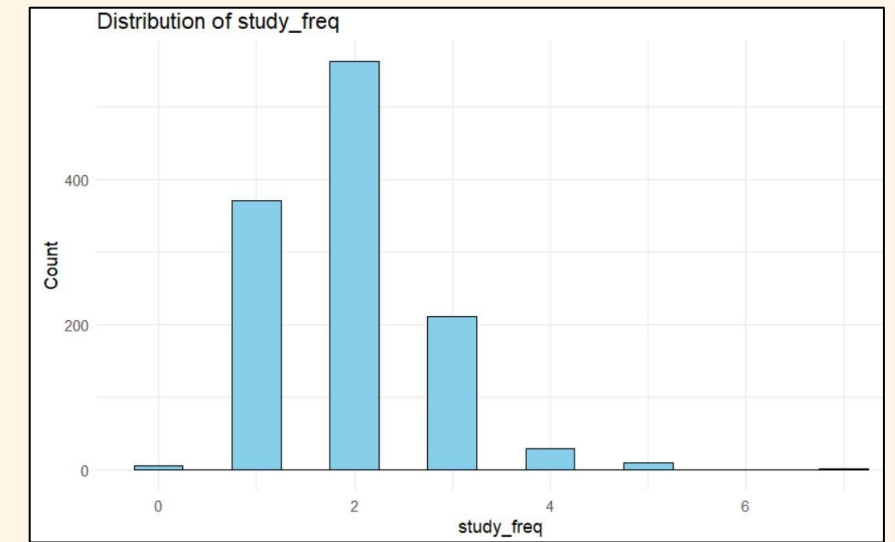
Students study between 1-4 hours per day

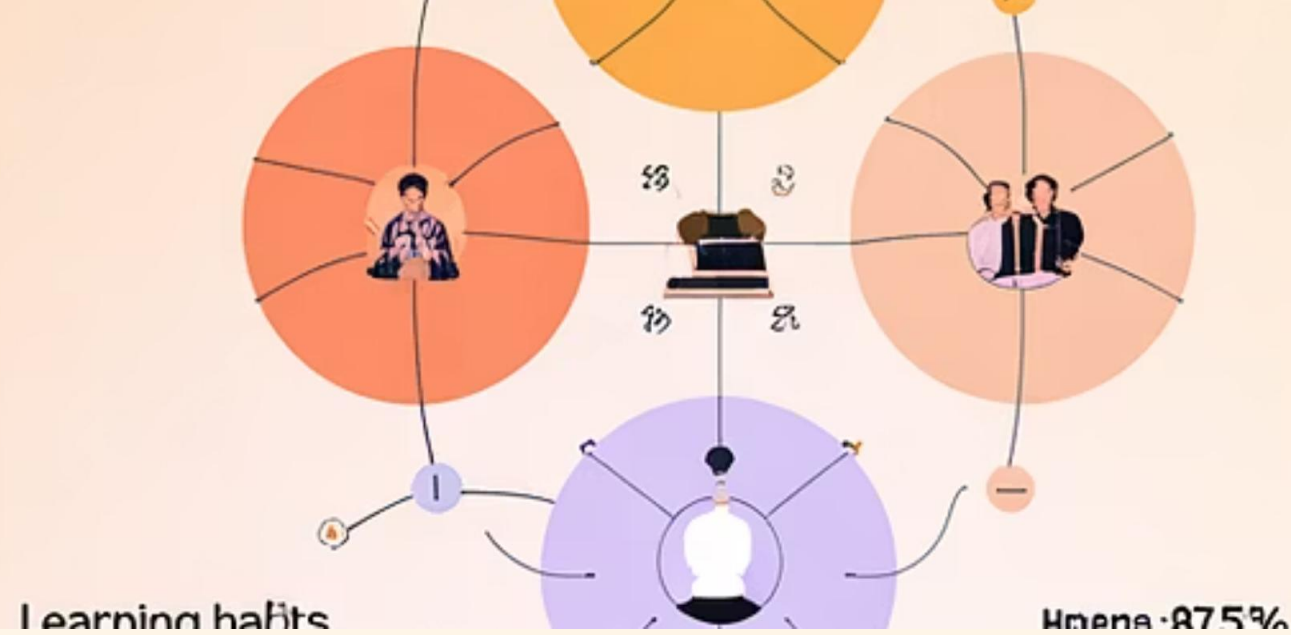


95%

Study Frequency

Report study frequency of 1-3 times





Findings



Verified University Records

Found to be the **best predictors** of final grades



Demographic Data

Found to be marginal predictors that **cannot be used alone for prediction**



Learning Habits Data

Found to be **very weak predictors** of final grades in our data

Key Insights & Conclusions

Research Questions 1 & 2

Confirmed prior findings: Academic metrics were strong predictors of CGPA; demographics alone had little to no predictive power.

Research Question 3

Accurate grade prediction based on learning habits requires **verified data** and **quality-related** metrics, **self-reports alone are not enough.**

