

MACHINE LEARNING

STUDENT PERFORMANCE PREDICTION

BY:

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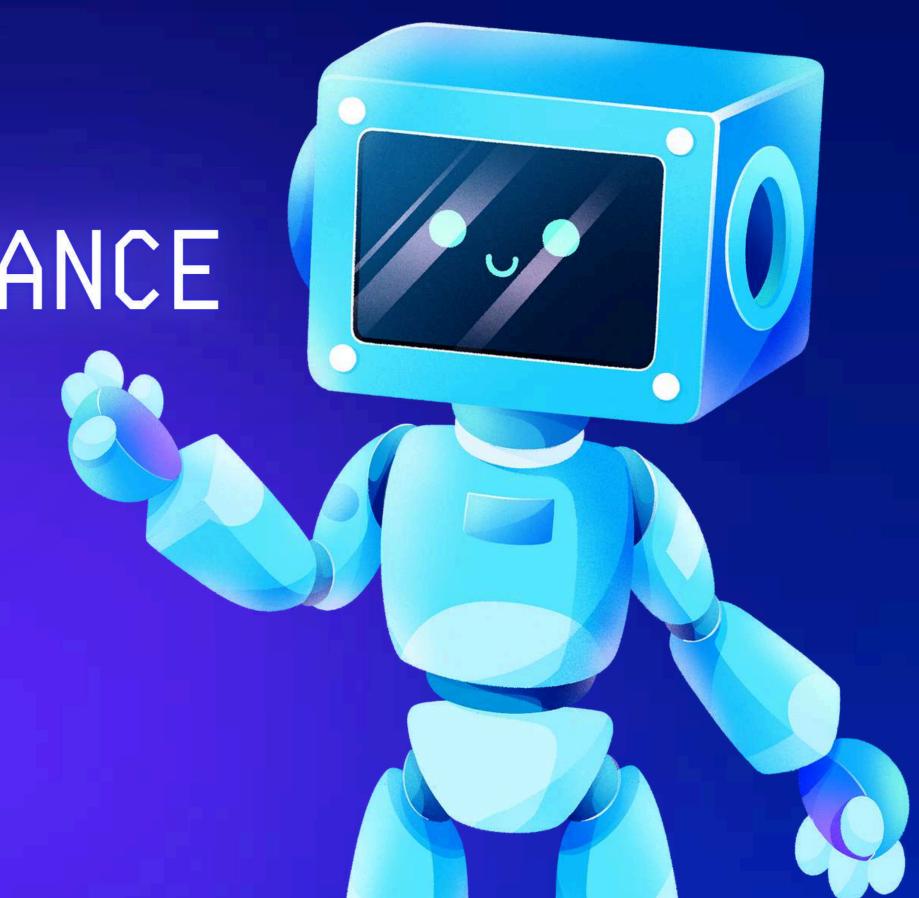




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INTRODUCTION

In the field of education, understanding and improving student performance is effective for teaching and learning.

- The advent of advanced technologies, particularly in the realm of machine learning, educators now have a powerful tool at their disposal.
- Our project endeavors to harness the potential of machine learning algorithms to predict student performance, thereby providing educators with valuable insights to tailor their teaching strategies, identify at-risk students, and intervene proactively to ensure academic success for all.





DATA SET-STUDENT PERFORMANCE



The Data Set for Student Performance was taken from kaggle:

https://www.kaggle.com/datasets/larsen0966/studentperformance-data-set

This data set talks about student achievement in secondary education of two Portuguese schools. The data attributes include student grades, demographic, social and school related features and it was collected by using school reports and questionnaires.



ATTRIBUTES IN DATA SET

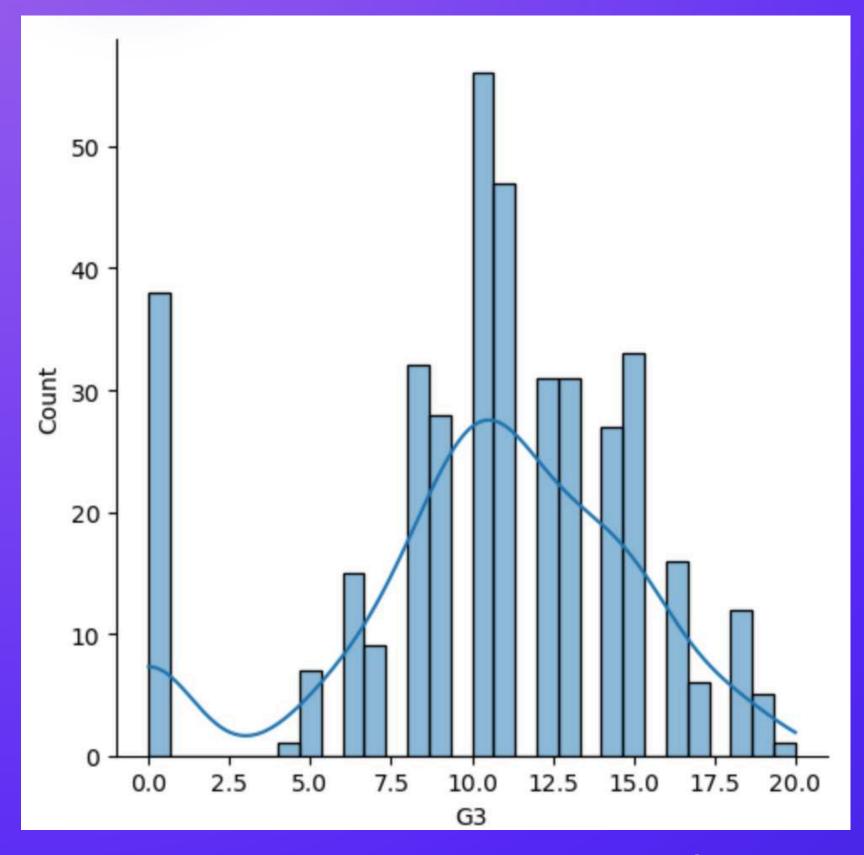
This data set consists of 33 columns which include the following:

- school, sex, age, address These attributes deal with basic features of a student.
- Famsize, Pstatus, Medu, Fedu, Mjob, Fjob, reason, guardian These attributes deal with the family of the student.
- Traveltime, studytime, failures, schoolsup, famsup, paid- These attributes deal with payement and study patterns of a student.
- nursery, higher, internet These attributes deal with the school education
- famrel, freetime, goout, activities These attributes deal with the recreational activities of a student.
- health, absences -These attributes deal with the health of student.
- G1,G2,G3 These attributes deal with the grades of the student.

TARGET VARIABLE: G3 (Final Grade of the Student)

EDA

Helps you understand the given data set and figure out patterns within the data and rectify anomalies.

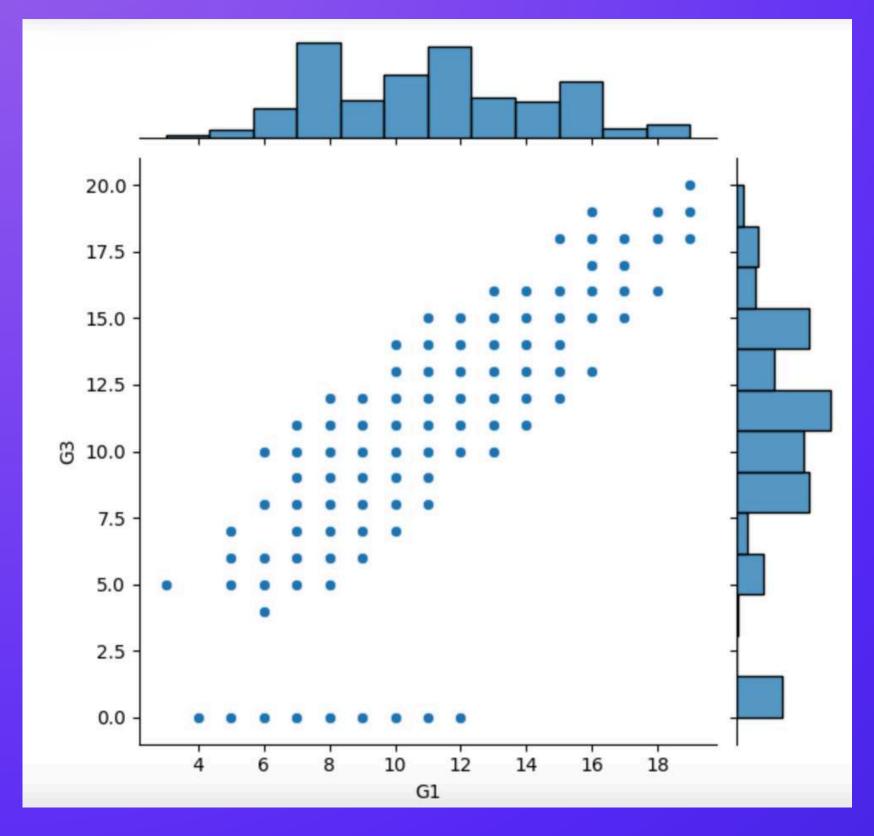


school age address - 0.8 famsize Medu Fedu traveltime - 0.6 studytime failures famsup - 0.4 activities nursery higher internet 0.2 famrel freetime goout - 0.0 Dalc Walc health absences - -0.2 G1 G2 sences G1 G2 G3 eetime goout Dalc health

this graph shows the distribution of target variable G3

This heatmap shows the correlation between different columns in the dataset

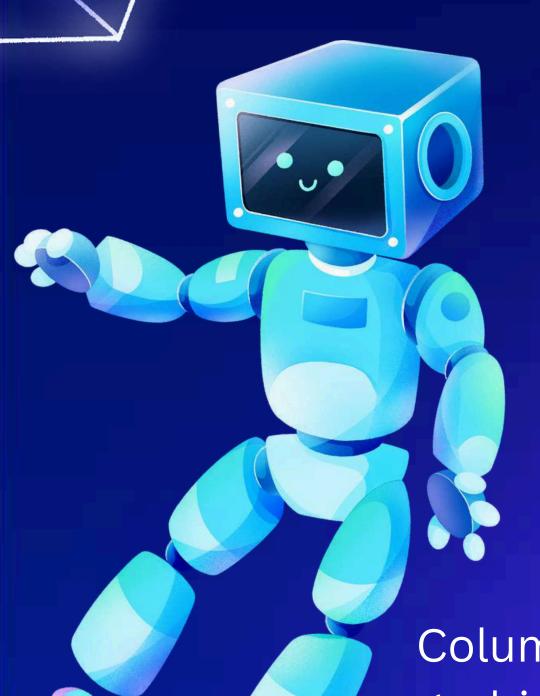
Also reveals missing data and allows you to visualize the data, making it easier to understand.



17.5 15.0 12.5 g 10.0 7.5 5.0 2.5 G1

Correlation between columns G1 and G3

correlation between G1 and G2



BINARY ENCODING

Process of converting non-numeric data to numeric form as machine learning algorithms ususally require numeric data.

```
studentperf['internet'] = studentperf['internet'].map({'yes': 1, 'no': 0})
studentperf['higher'] = studentperf['higher'].map({'yes': 1, 'no': 0})
studentperf['nursery'] = studentperf['nursery'].map({'yes': 1, 'no': 0})
studentperf['activities'] = studentperf['activities'].map({'yes': 1, 'no': 0})
studentperf['famsup'] = studentperf['famsup'].map({'yes': 1, 'no': 0})
studentperf['school'] = studentperf['school'].map({'GP': 1, 'MS': 0})
studentperf['sex'] = studentperf['sex'].map({'M': 1, 'F': 0})
studentperf['address'] = studentperf['address'].map({'U': 1, 'R': 0})
studentperf['famsize'] = studentperf['famsize'].map({'GT3': 1, 'LE3': 0})
```

Columns like sex that have values male and female can be mapped to binary values 1 and 0 to make it understandable to the model. Here since there are only 2 unique values for each column we can use binary encoding but if there are more unique values we can use either label encoding or one-hot encoding.

REMOVING UNNECESSARY COLUMNS

Important to remove irrelevant information from dataset as it may affect the performance of the model, and it will result in faster training time.

```
studentperf.pop("reason")
studentperf.pop("guardian")
studentperf.pop("schoolsup")
studentperf.pop("paid")
studentperf.pop("romantic")
studentperf.pop("Pstatus")
studentperf.pop("Mjob")
studentperf.pop("Fjob")
```

MODEL SELECTION

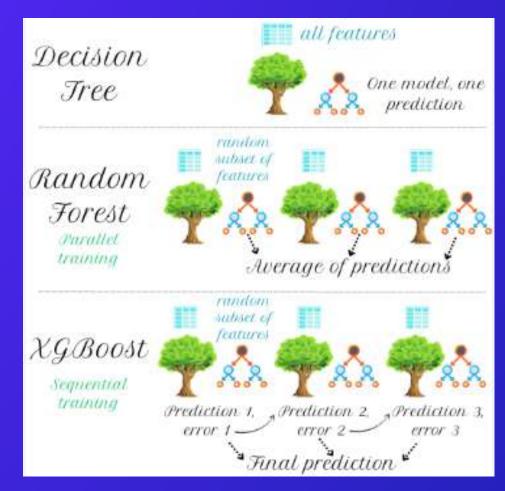


There are many Machine learning algorithms available but we should select a model that gives a high accuracy in determining the target variable.

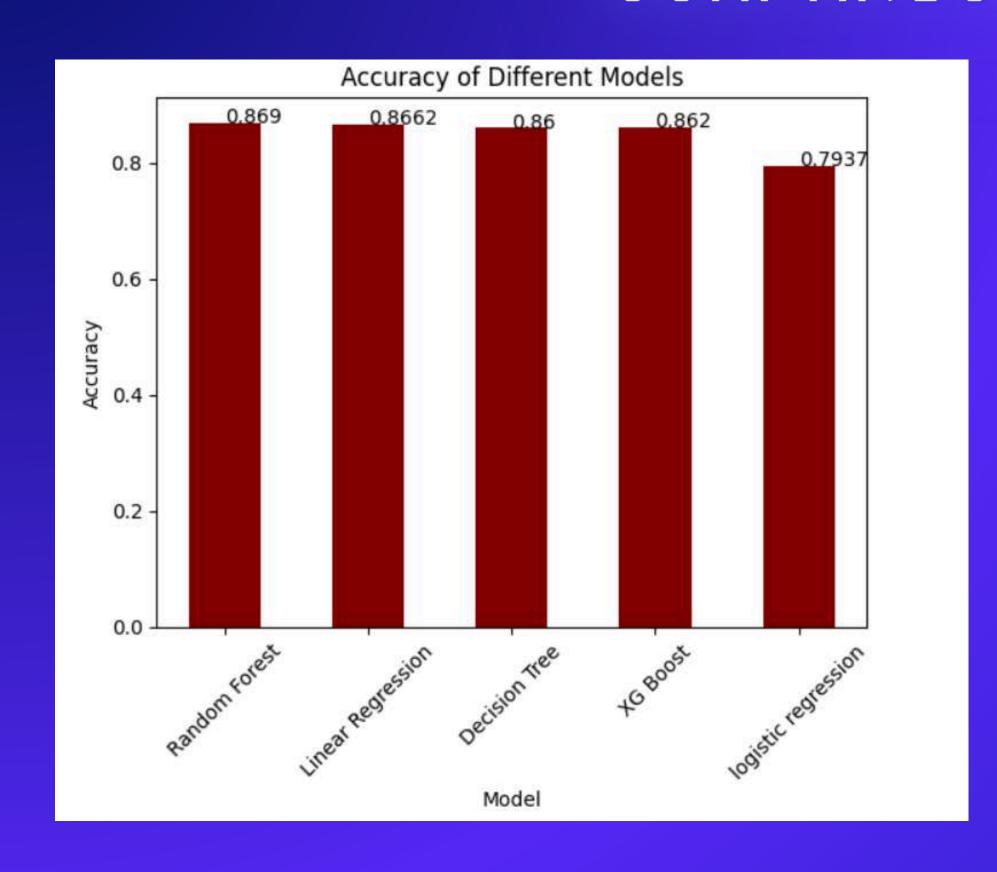
02

We tested 5 different Machine learning models and compared their R2 scores to determine the most effective model amongst them. These models include:

- Linear Regression
- Logistic Regression
- Decision Tree
- Random Forest
- XG Boost



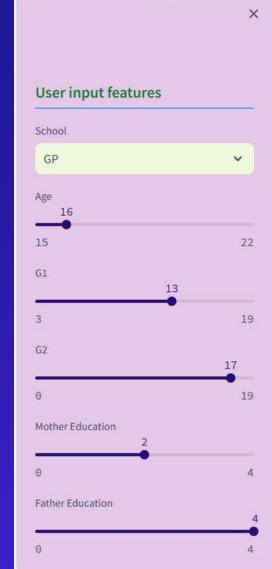
COMPARISON

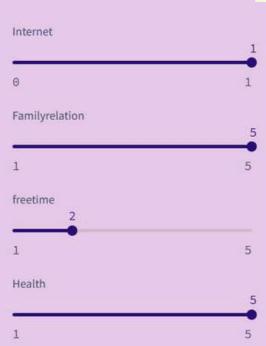


As seen from the figure, Random Forest has the most accuracy in comparison to other models and logistic regression has the least accuracy. Hence, the best model to work on the data set of Student Performance Prediction is Random Forest

GITHUB LINK: https://github.com/zobiyaFathima/zobiya-codes

STREAMLIT





STUDENT PERFORMANCE PREDICTION APP

This app predicts the performance of a student based on various features pertaining to a student. This app finds the final grade G3 of a student.

Data set obtained from kaggle:

https://www.kaggle.com/datasets/larsen0966/studentperformance dataset



User input Features

	ltime	studytime	failures	activities	internet	famrel	freetime	health	absences	G1	G2
0	1	4	1	1	1	5	2	5	10	13	1

Predicted Student Score-G3

value 17.6



