

PROACTIVE STUDENT SUPPORT AND NEXT BEST ACTION

Group 25



CLIENT

The University of
Melbourne



**Prof. Kate Smith Miles
&
Mr. Milad Chenaghlou**

TEAM



Milad Chenaghlou
Supervisor



Sudheer Kumar
1051717



Haopeng Yan
962332



Xiaowen Jin
1023499



Yiming Xu
863672



Harshal Shah
1020849

GROUP 25

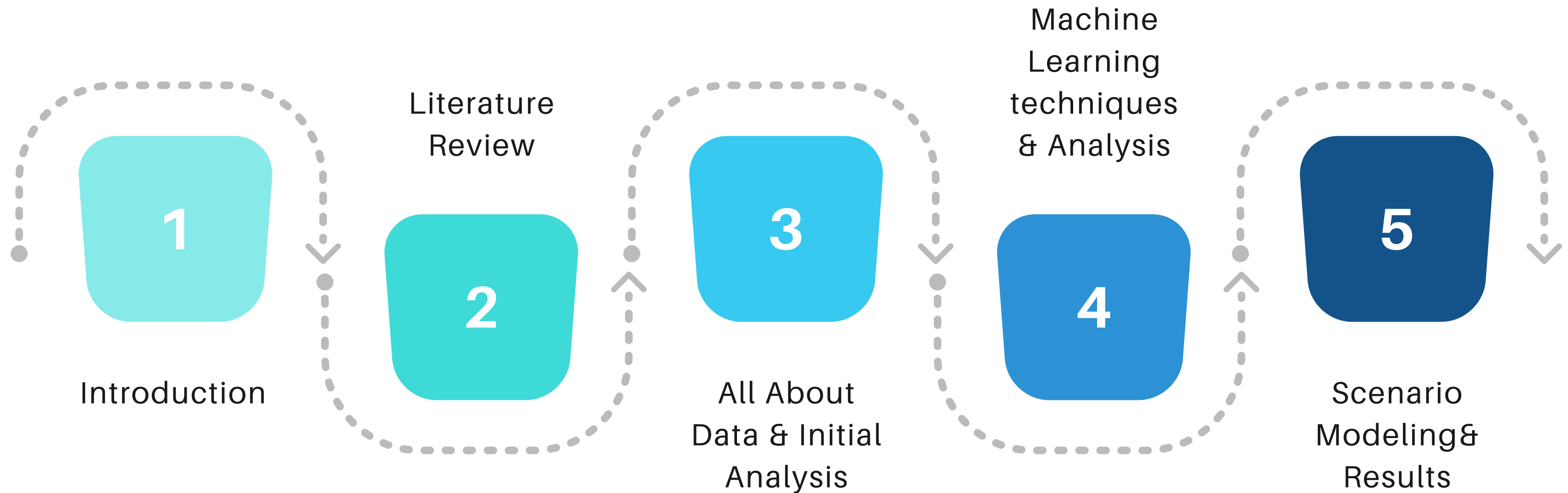
Overview of Roles

<div>Sudheer Kumar Kolla</div> <div>Haopeng Yan</div> <div>Xiaowen Jin</div> <div>Yiming Xu</div> <div>Harshal Shah</div>		Data Extrapolation	Regression Models	Scenario Modeling
		Initial Data Analysis	Classification For Demographic Data	Scenario Modeling Story Line
		Classification For Demographic Data	Hyperparamater Tuning & Results For Demographic Data	Linear Discriminant Analysis On Demographic Data
		Initial Data Analysis	Clustering Research	Scenario Modeling Story Line
		Classification & Regression Models (For Entire Dataset)	Hyperparameter Tuning & Results For Entire Dataset	Linear Discriminant Analysis On Entire Dataset

All the team members contributed equally towards report and presentation

PROACTIVE STUDENT SUPPORT AND NEXT BEST ACTION

Project Flow



INTRODUCTION



CLIENT

The University of
Melbourne



DOMAIN

Student Success
Prediction &
Recommendation



GOAL

Empowering the
educational institutions
to provide a better
support to their students



SCOPE

Identify the students at
risk and accordingly,
reach out to them and
provide relevant
suggestions

LITERATURE SURVEY

Identifying At-Risk Students: Using Machine Learning Techniques:
A Case Study with IS 100

1

Dimensions of student success: a framework for defining and
evaluating support for learning in higher education

2

Connect for Success: A proactive student identification and
support program

3

Analyzing and Predicting Student's Performance by Means of
Machine Learning: A review

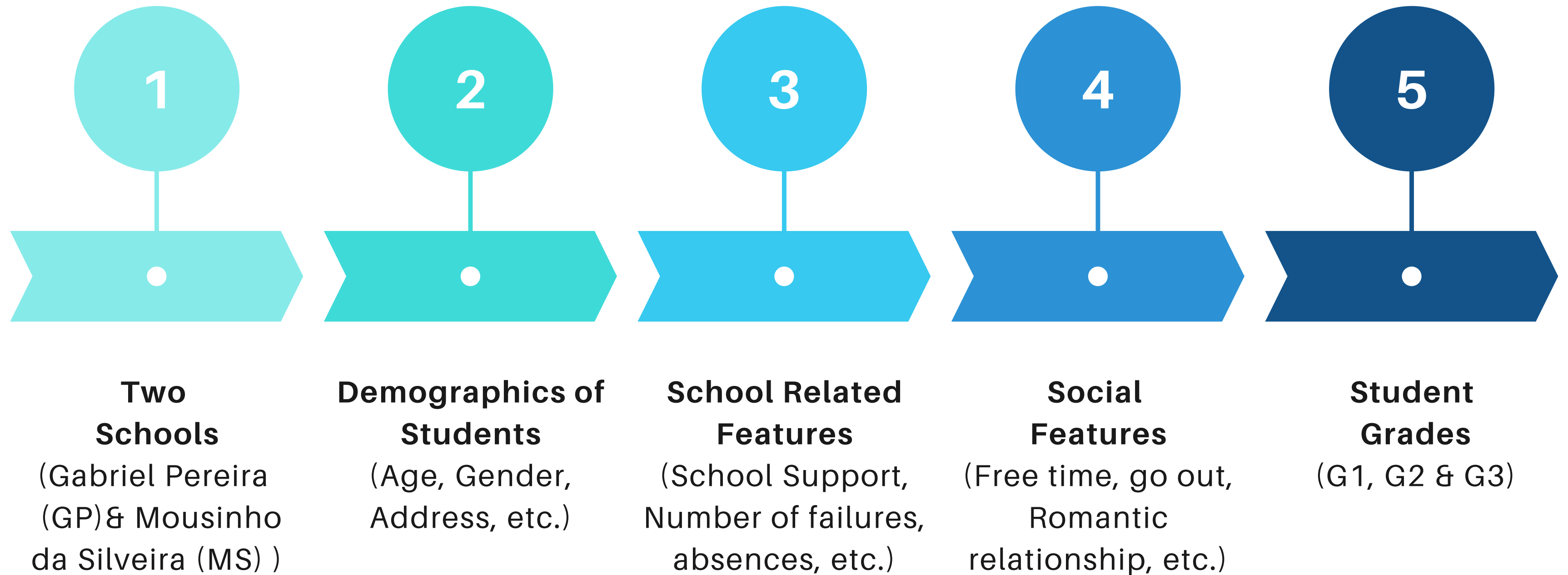
4

Identifying students at risk of
academic failure within the educational data mining framework

5

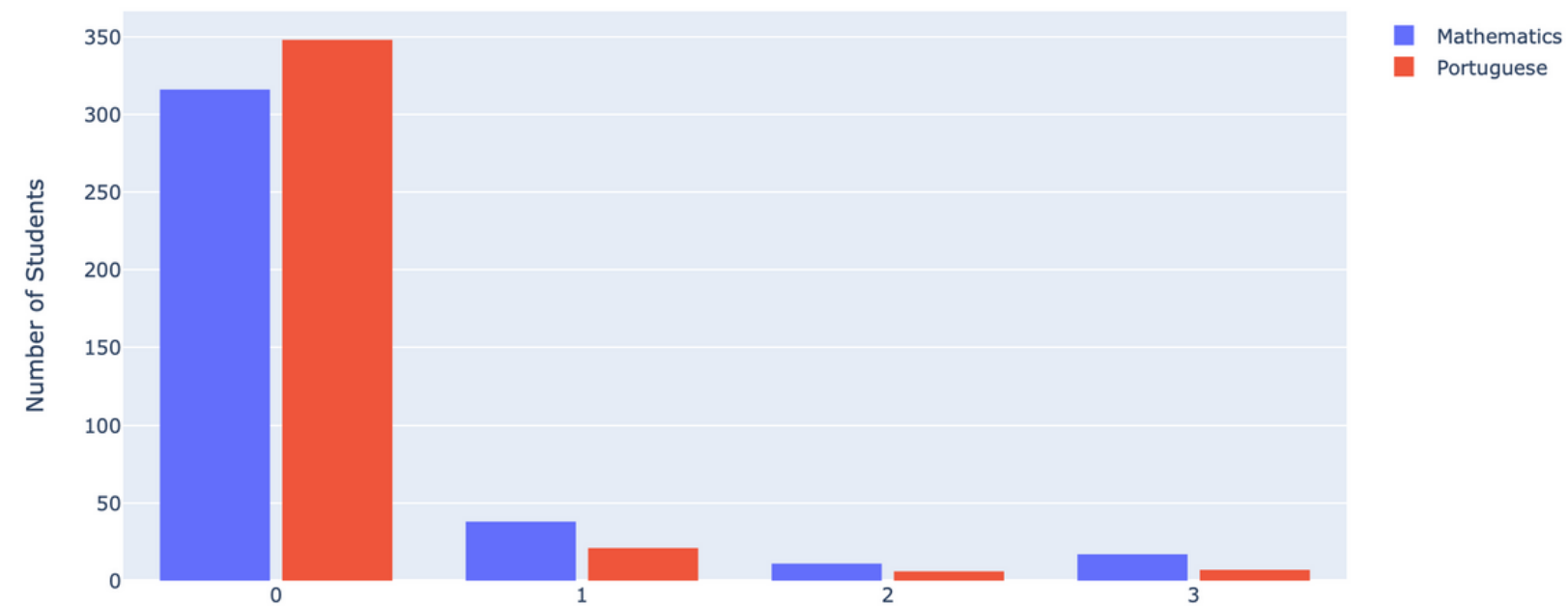
DATASET INFORMATION

UCI Machine Learning Repository

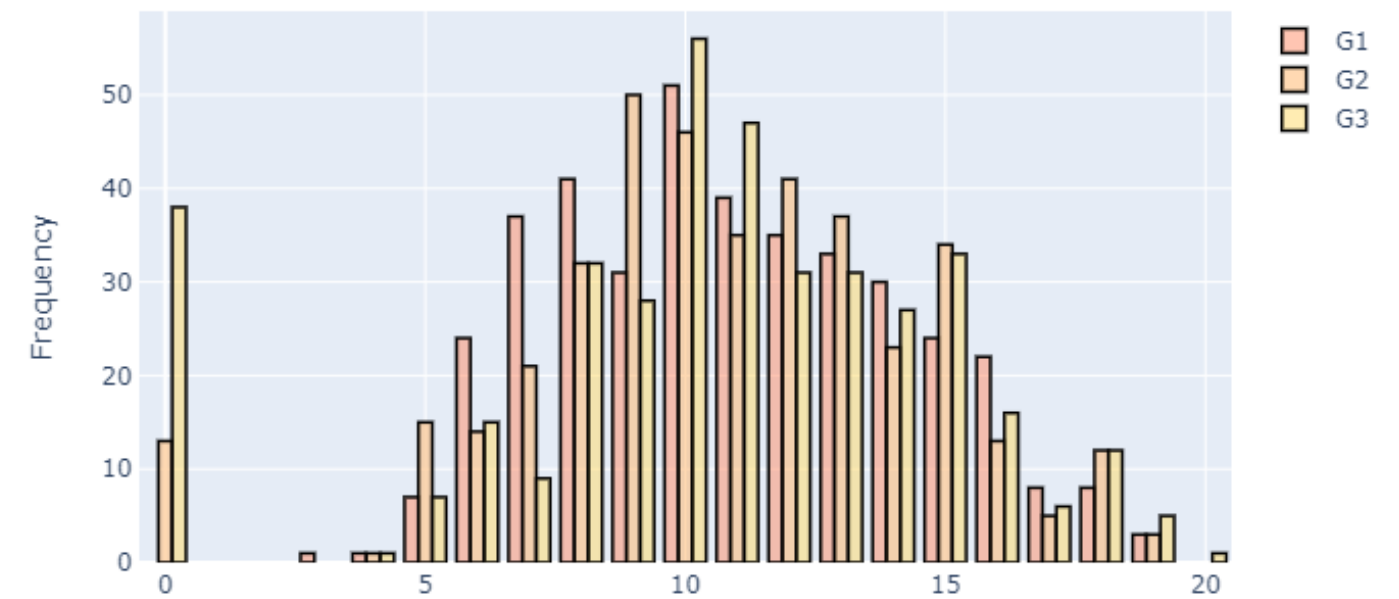


In Total, there are 33 features varying from school to student grades and for two subjects, Mathematics and Portuguese

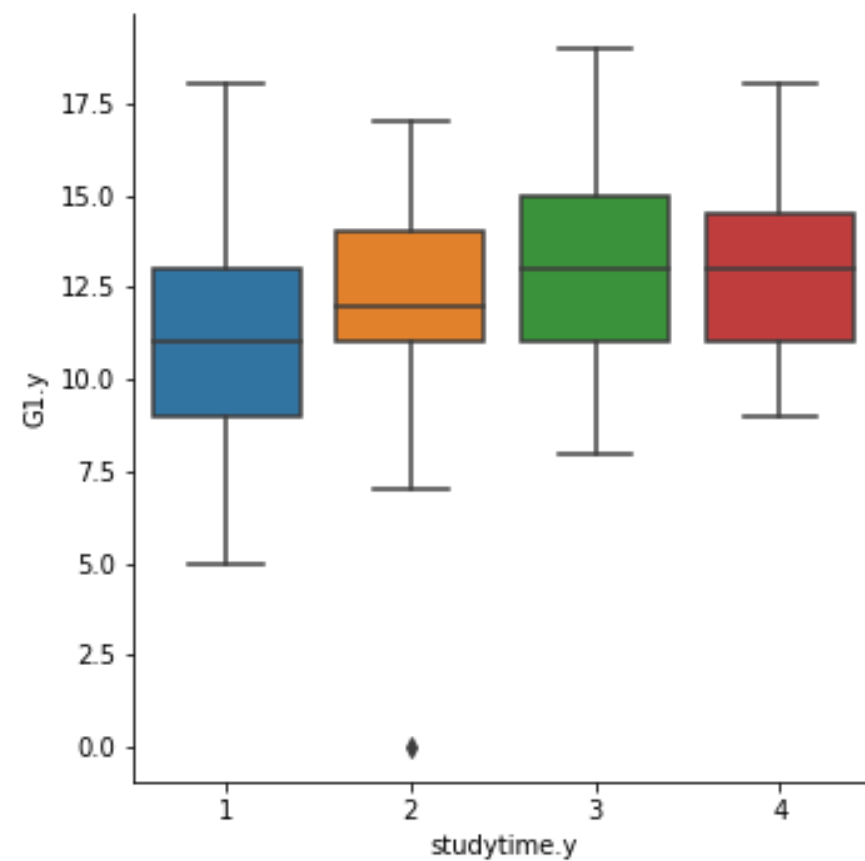
EXPLORATORY DATA ANALYSIS



Number of Failures for Maths and Portuguese



Portuguese Grades Distribution

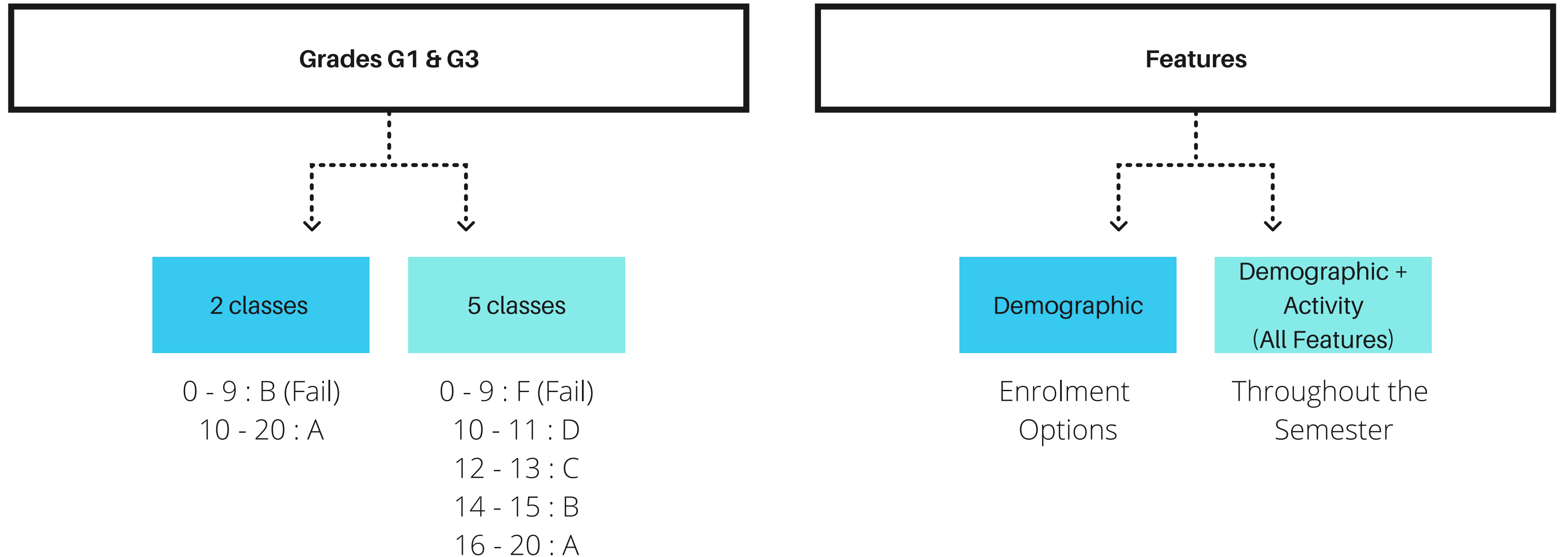


Study Time vs G1 Grade for Portuguese



Failures vs G3 Grade for Portuguese

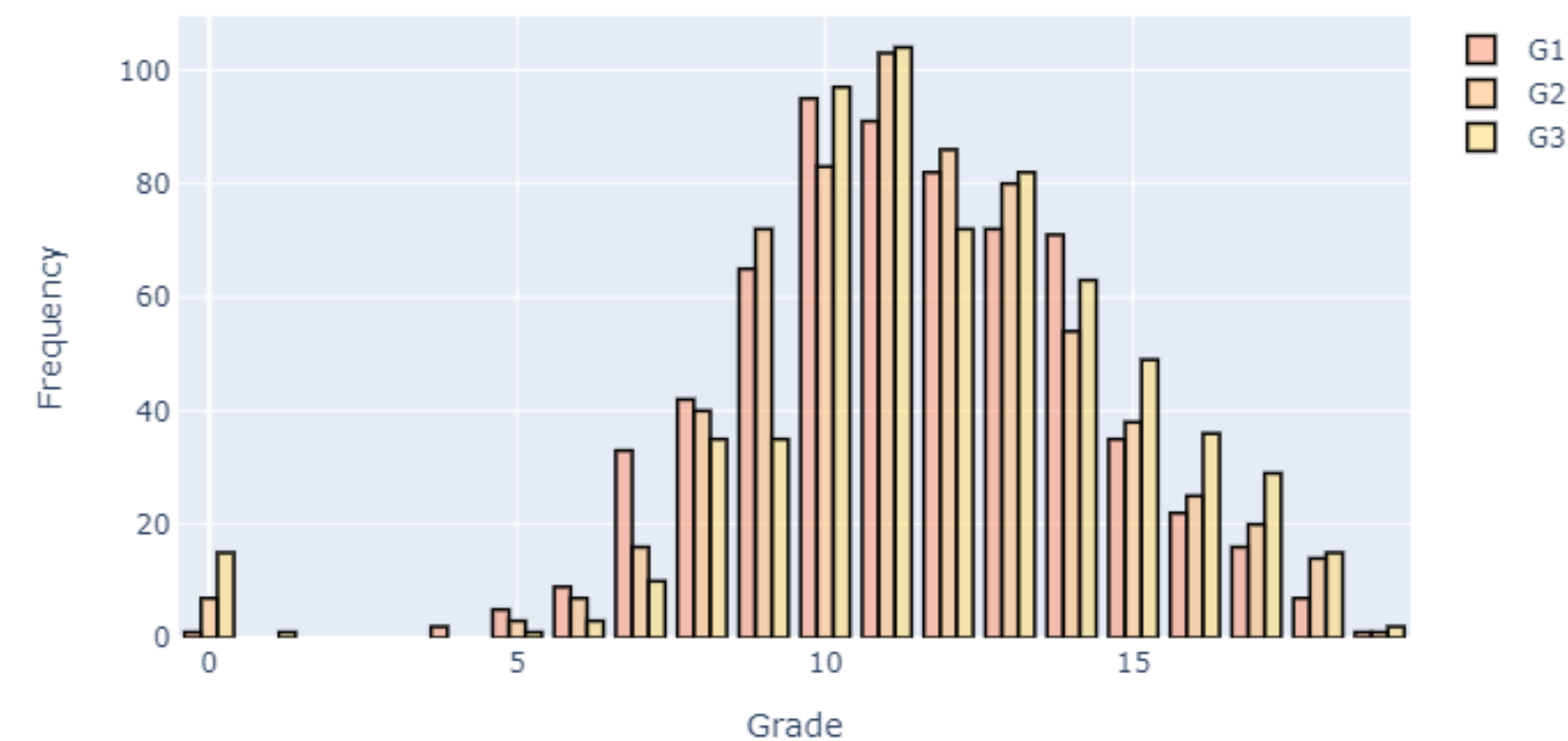
DIVISION OF FEATURES & GRADES INTO CLASSES



* Applied on all the datasets which are used

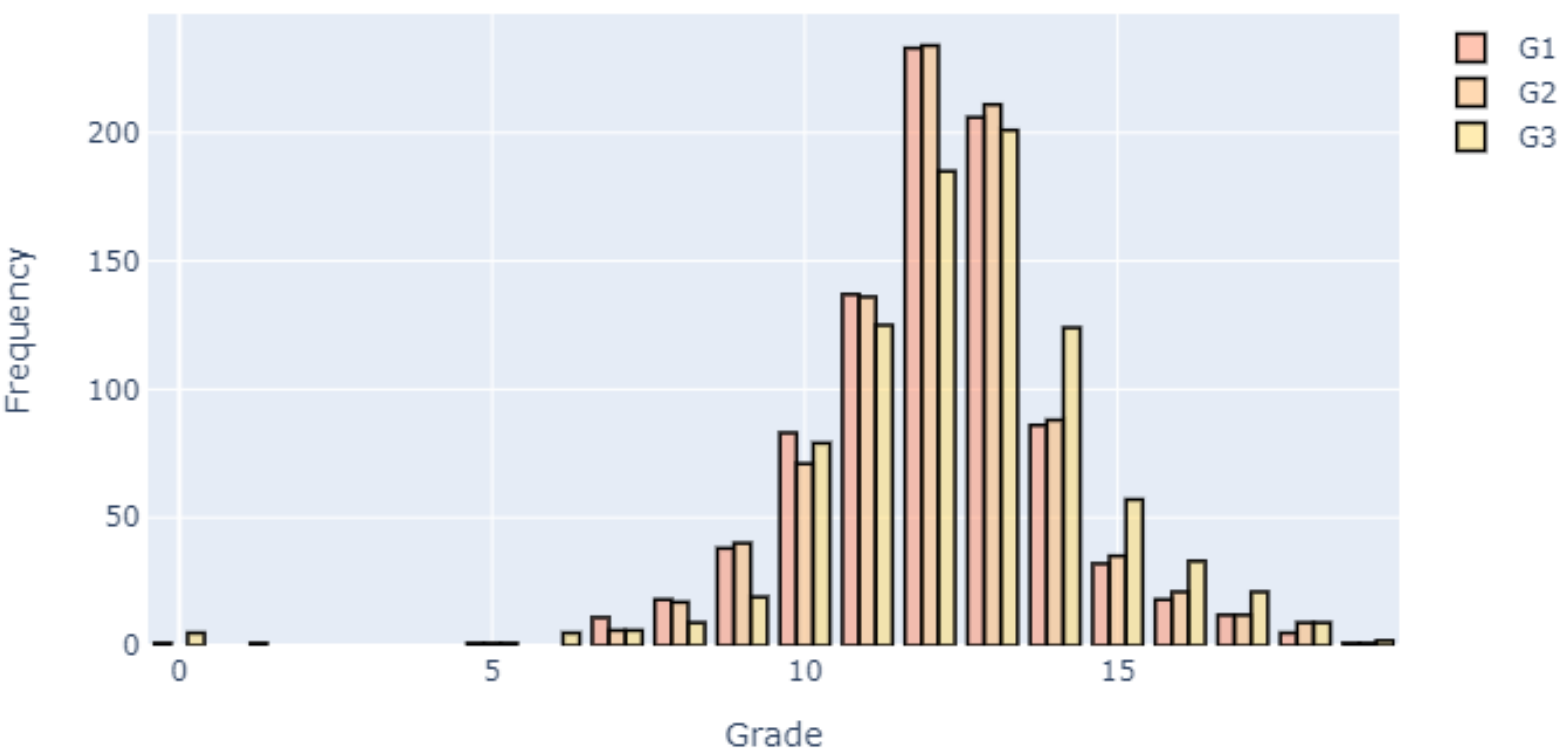
DATA EXTRAPOLATION

Distribution of G1, G2, G3 for Portuguese



Original Data
(382 rows)

Distribution of G1, G2, G3 for Portuguese

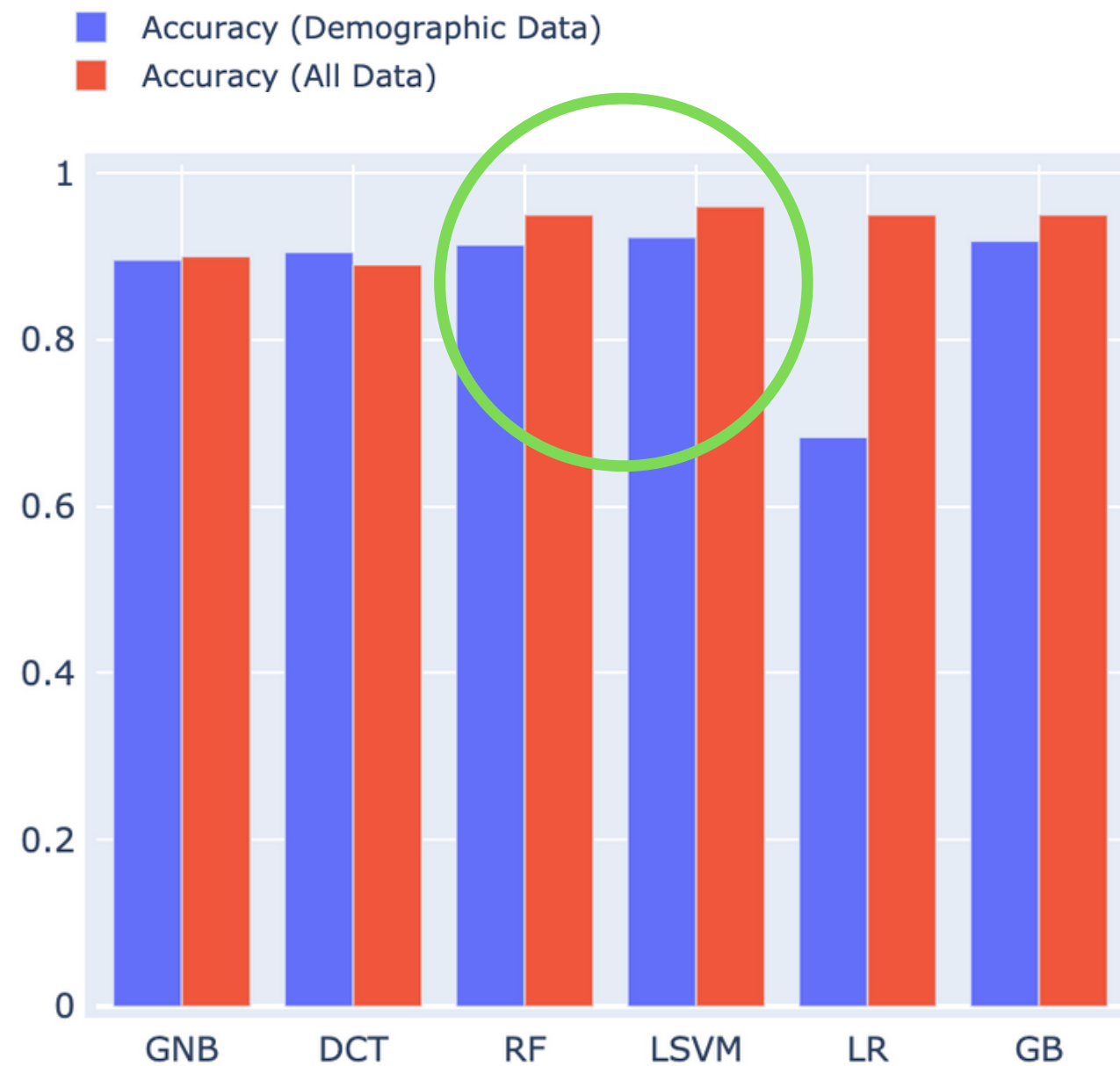


Extrapolated Data
(882 rows)

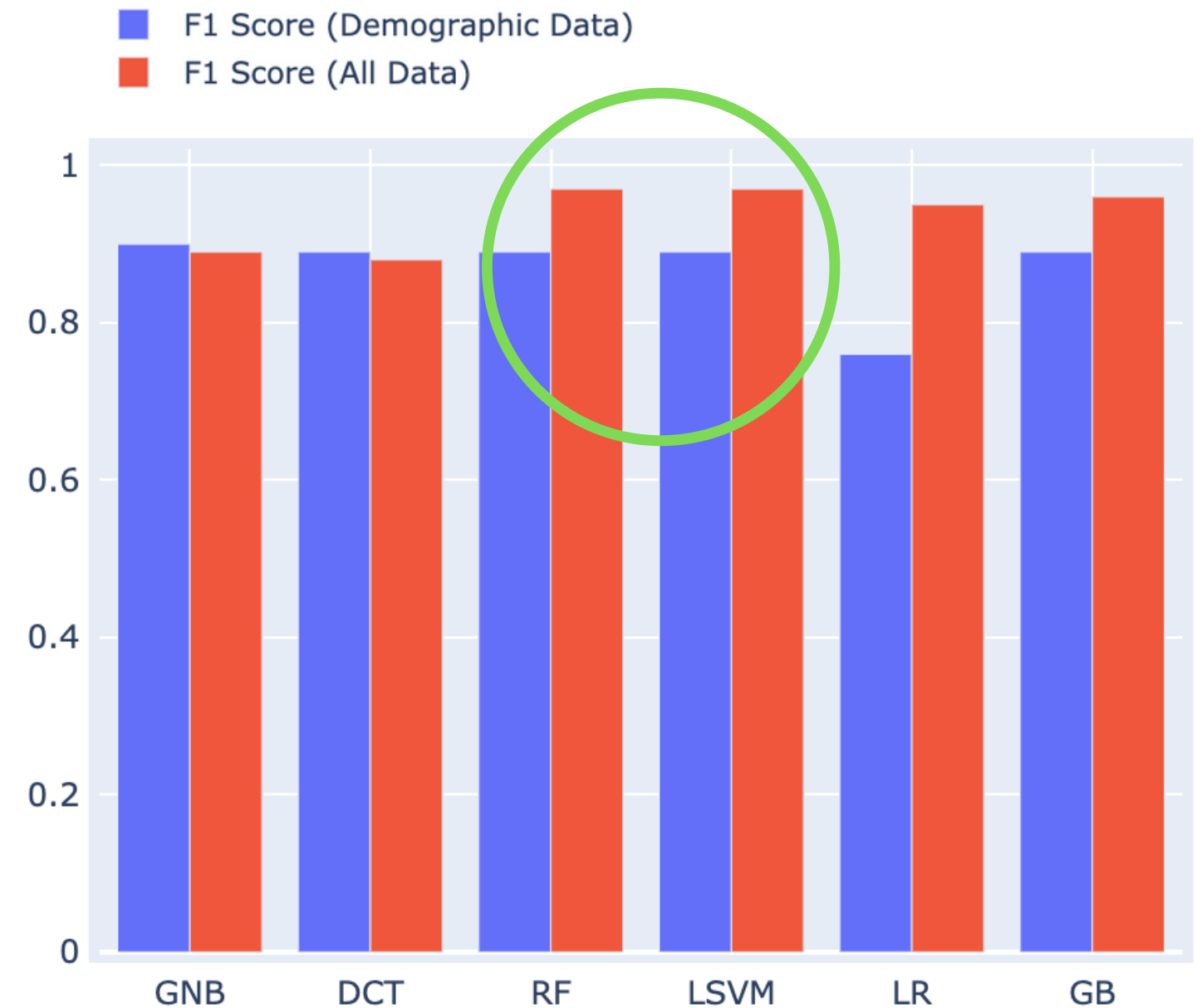
After comparing the results with original data, the extrapolated data gave better results and thus the extrapolated data was used for further analysis
(Performed for maths dataset also)

RESULTS OF PREDICTION FROM EXTRAPOLATED DATA

Prediction of G1 Score (Portuguese)
using Binary Classification Method



Accuracy



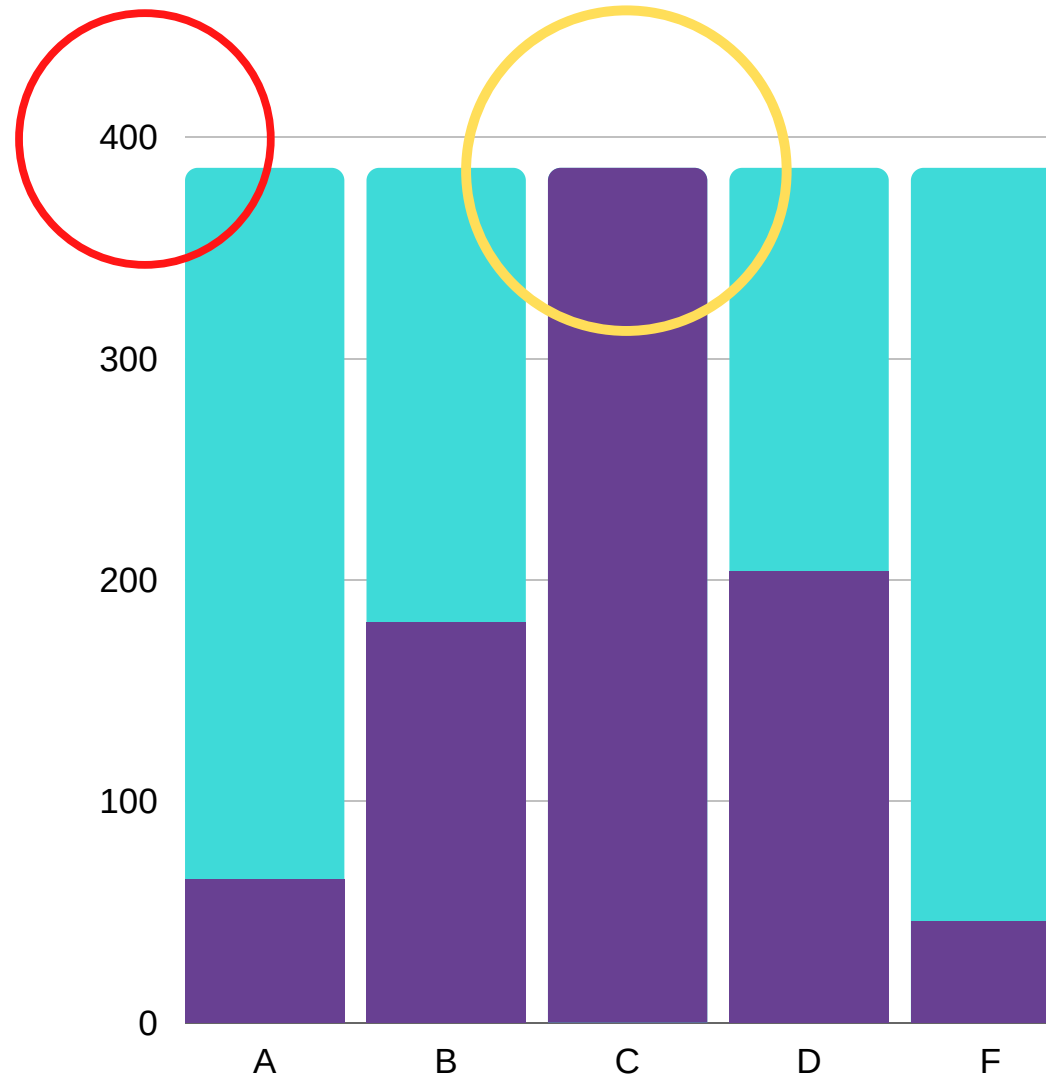
F1 Score

Results were obtained for Maths using Binary classification and for both the subjects, using 5 class classification technique

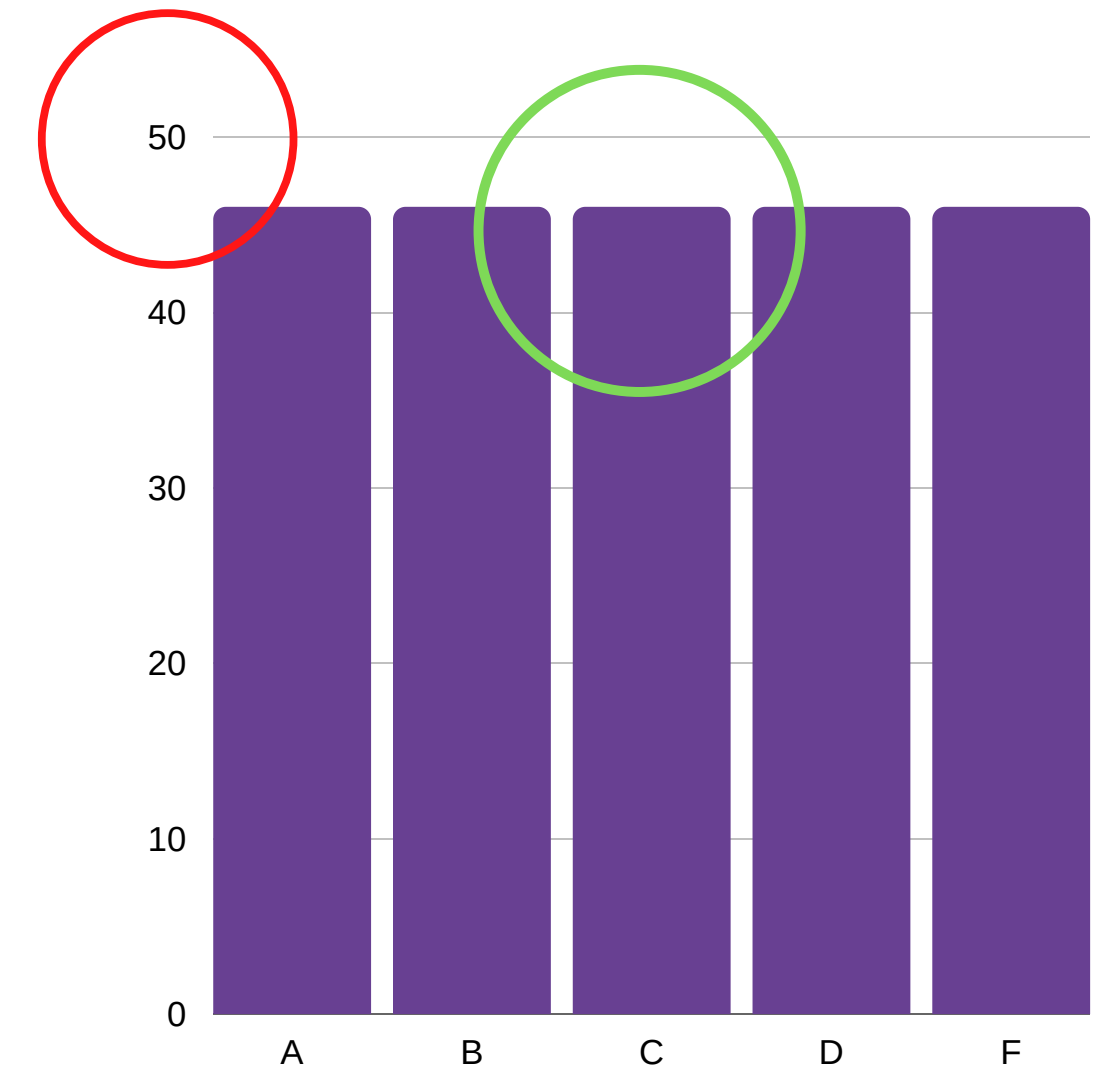
DATA UPSAMPLING & DOWNSAMPLING (HANDLE DATA IMBALANCE)



Extrapolated Distribution of the grade G3 for Portuguese data



Upsampled Data (Portuguese)

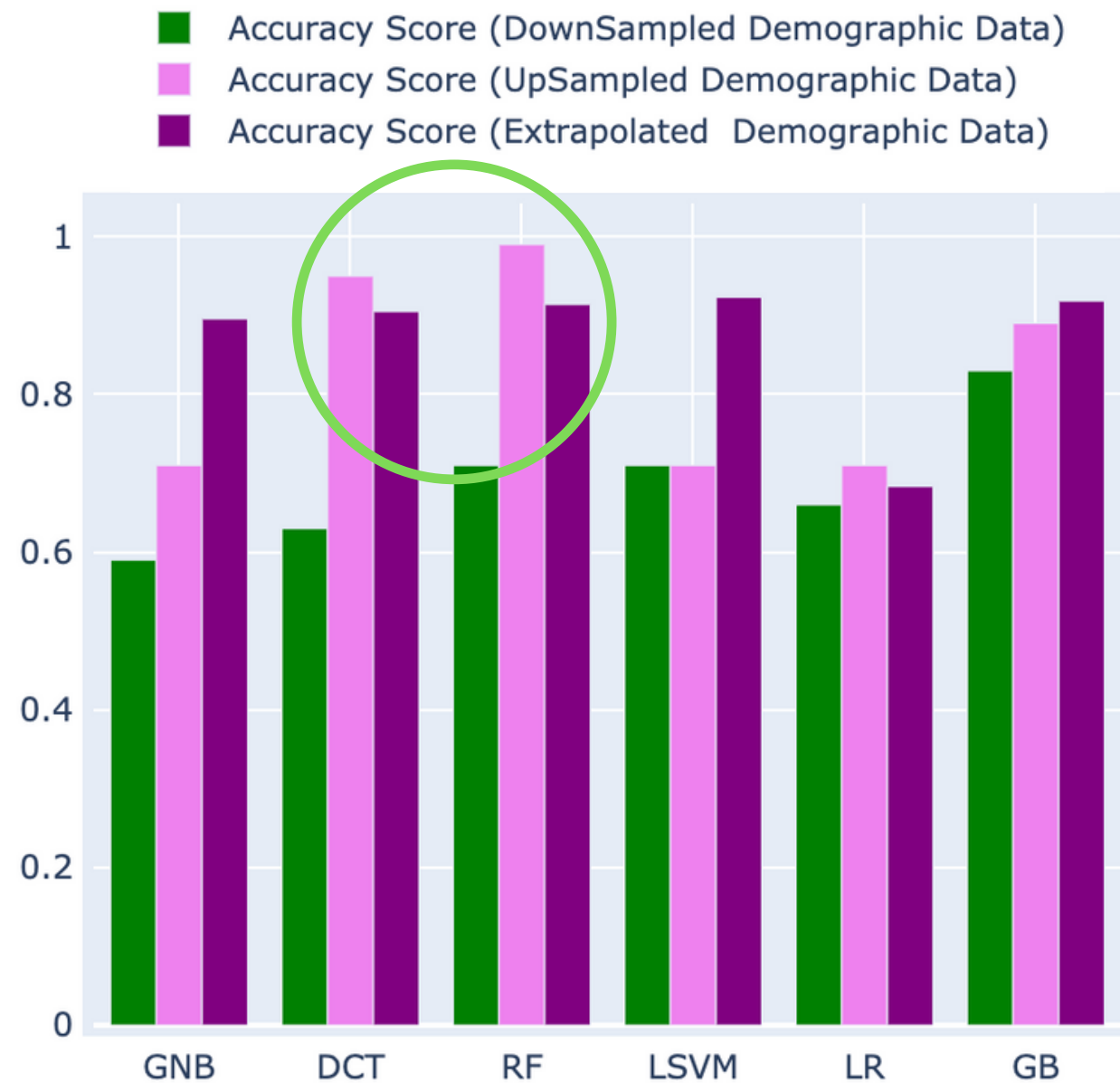


Downsampled Data (Portuguese)

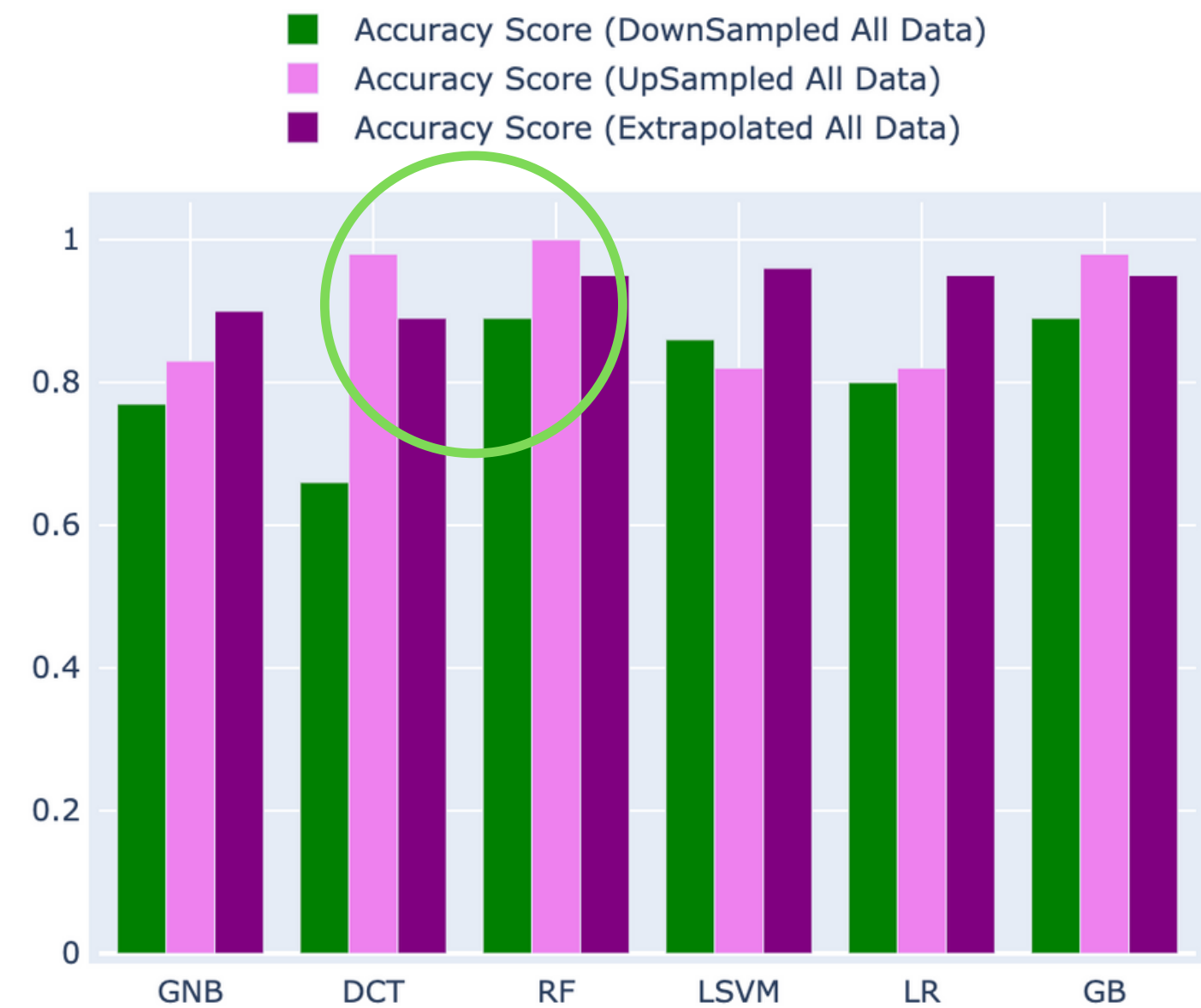
Similar Sampling techniques were performed on the data used for Binary Classification (Both Maths and Portuguese) and also for Maths data in 5 class classification

RESULTS OF PREDICTION USING SAMPLED DATA

Prediction of G1 Score (Portuguese) using Binary Classification Method



Comparison between accuracy obtained by 6 models



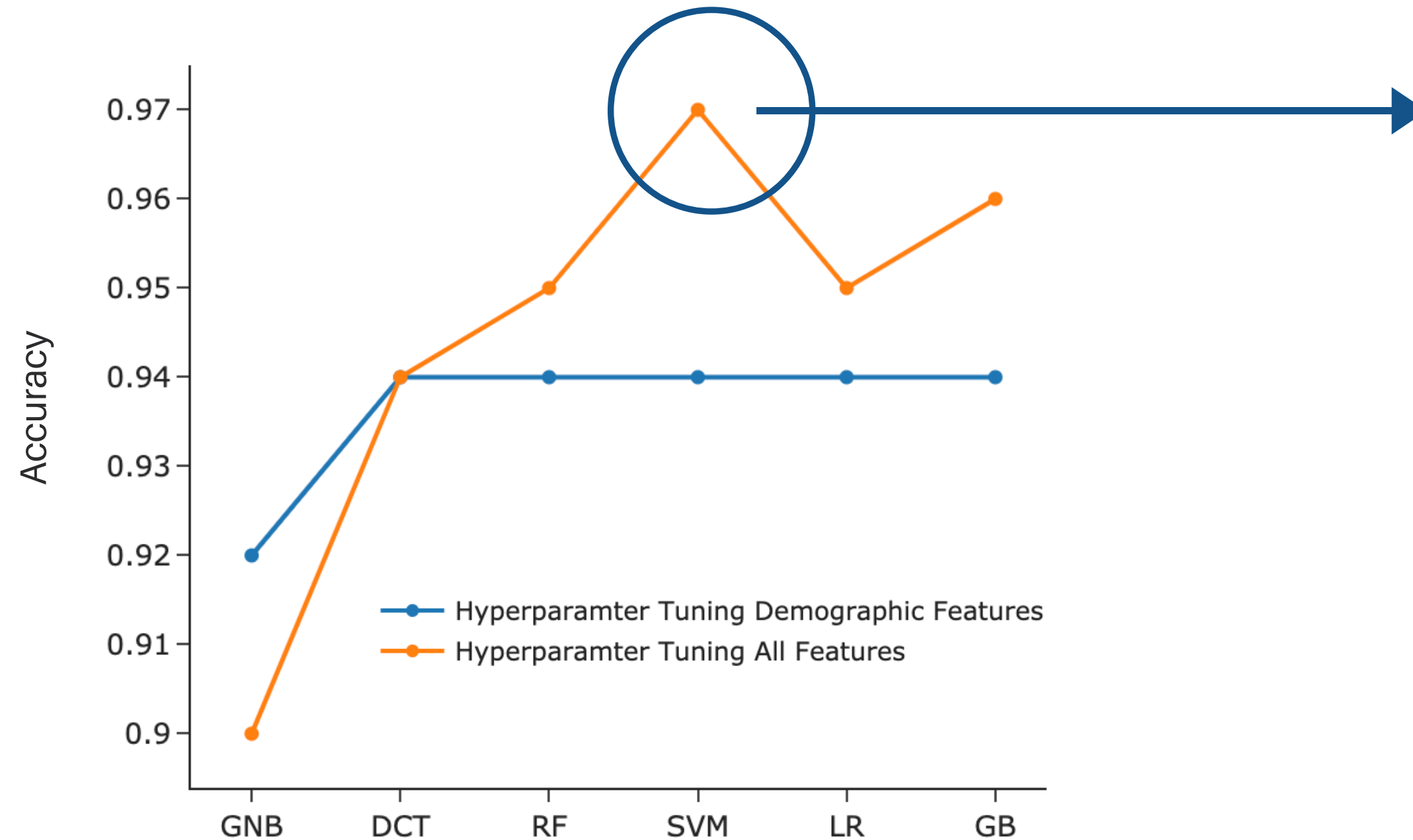
Comparison between accuracy obtained by 6 models

Results were obtained for Maths using Binary classification and for both the subjects, using 5 class classification technique

HYPERPARAMETER TUNING

GridSearchCV and RandomizedSearchCV

Prediction of G1 Score (Portuguese) using Binary Classification Method



Best Parameters obtained:

$C = 1$
Kernel = 'Linear'
degree = 3
gamma = 1

Results were obtained for Maths using Binary classification and for both the subjects, using 5 class classification technique
(Due to large computation time and memory issues, RandomizedSearchCV was preferred (Not much difference in results obtained by using GridSearchCV))

MOST IMPORTANT FEATURES

(Commonly found from all the above techniques)

Demographic Data

Mother's Education

Father's Job Status

Family Size

Age

School

Entire Dataset

Study Time

Father's Job

Go Out

Number of Absences

Number of Failures



PROACTIVE STUDENT SUPPORT AND NEXT BEST ACTION

SCENARIO MODELLING

Everything looks good with a story!

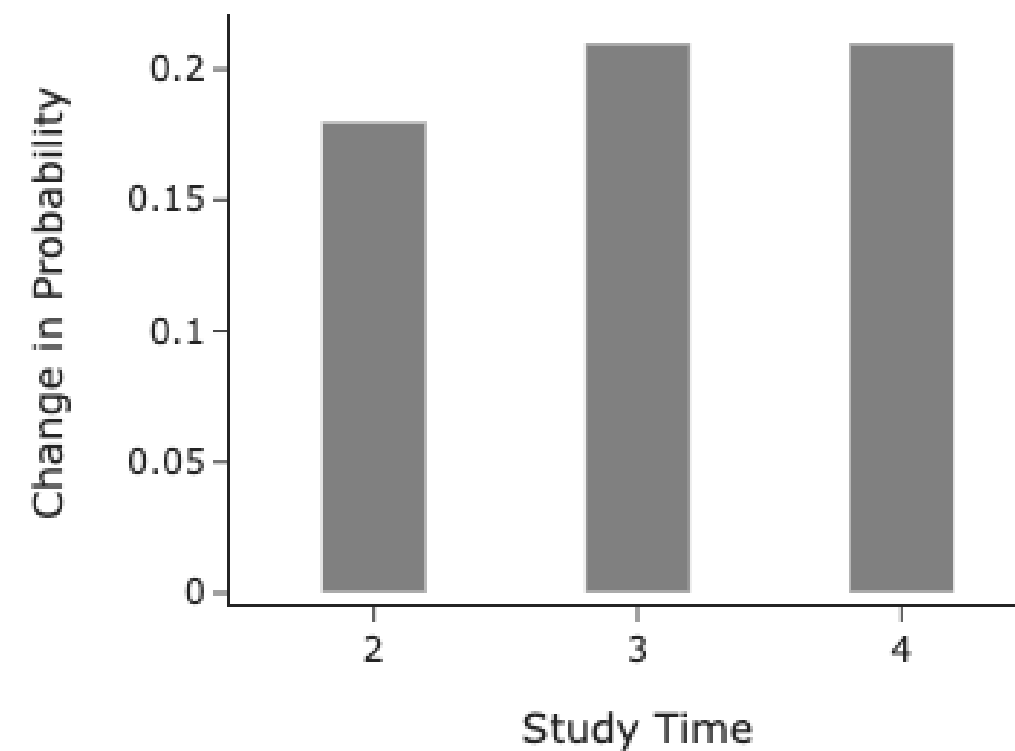
SCENARIO 1

Student ID - 342

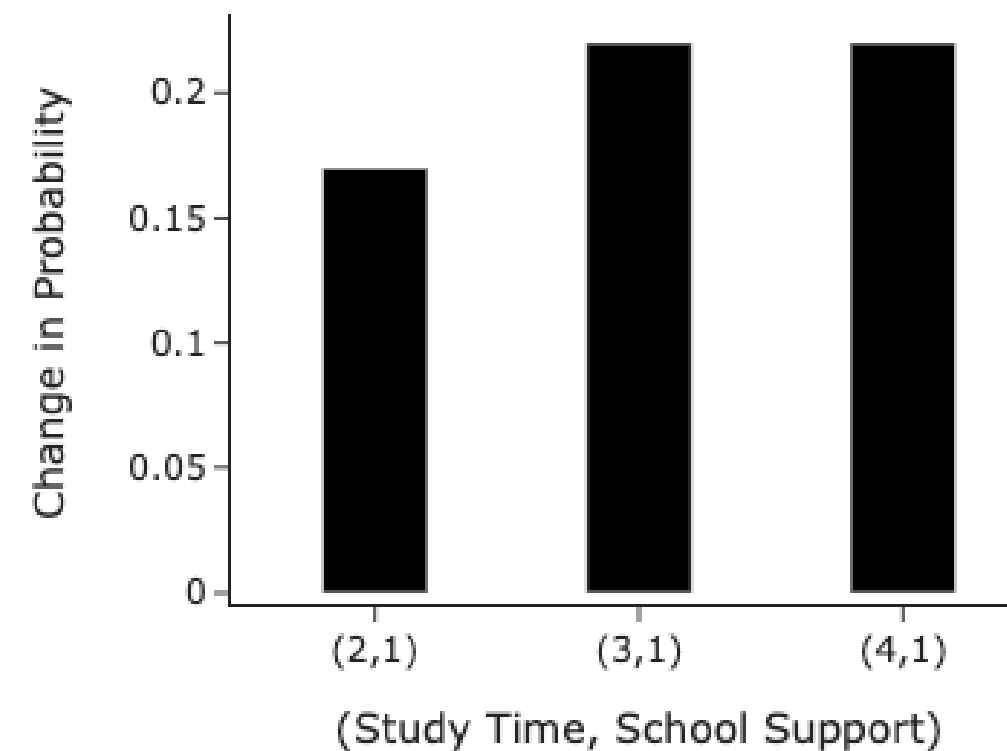
Initial probability of passing in G1 exam of Portuguese subject is 0.26

Initial study time value is 1 (studying less than 2 hours per week)

Scenario with Study Time



Scenario with Study Time & School Support



Similarly, more scenarios were explored using multiple features for both the subjects

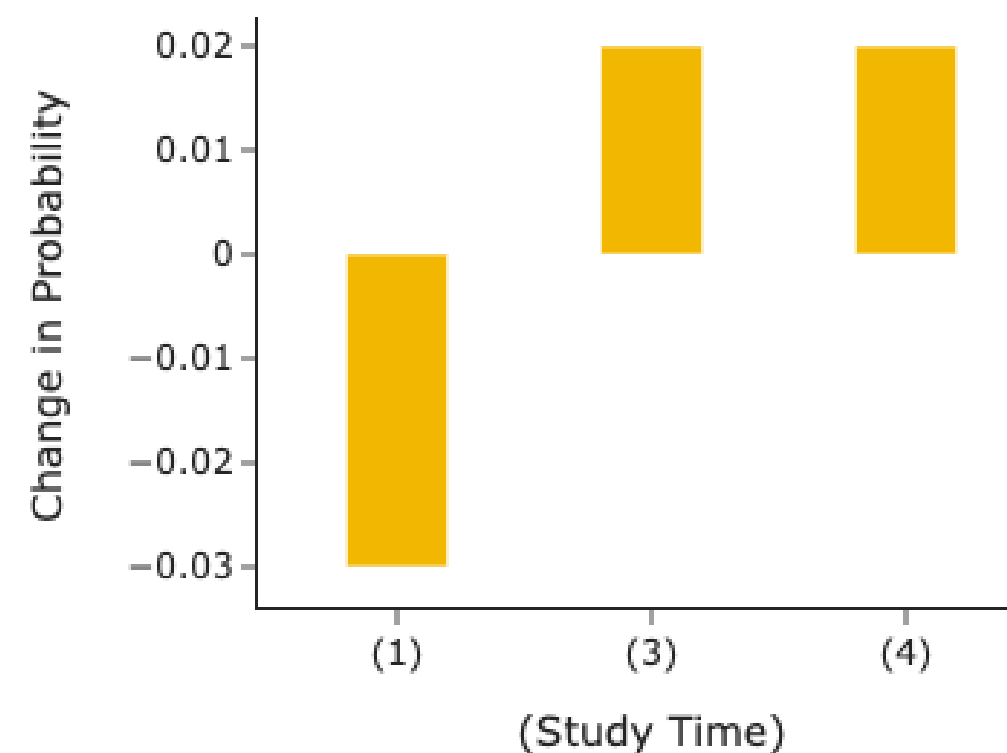
SCENARIO 2

Student ID - 45386

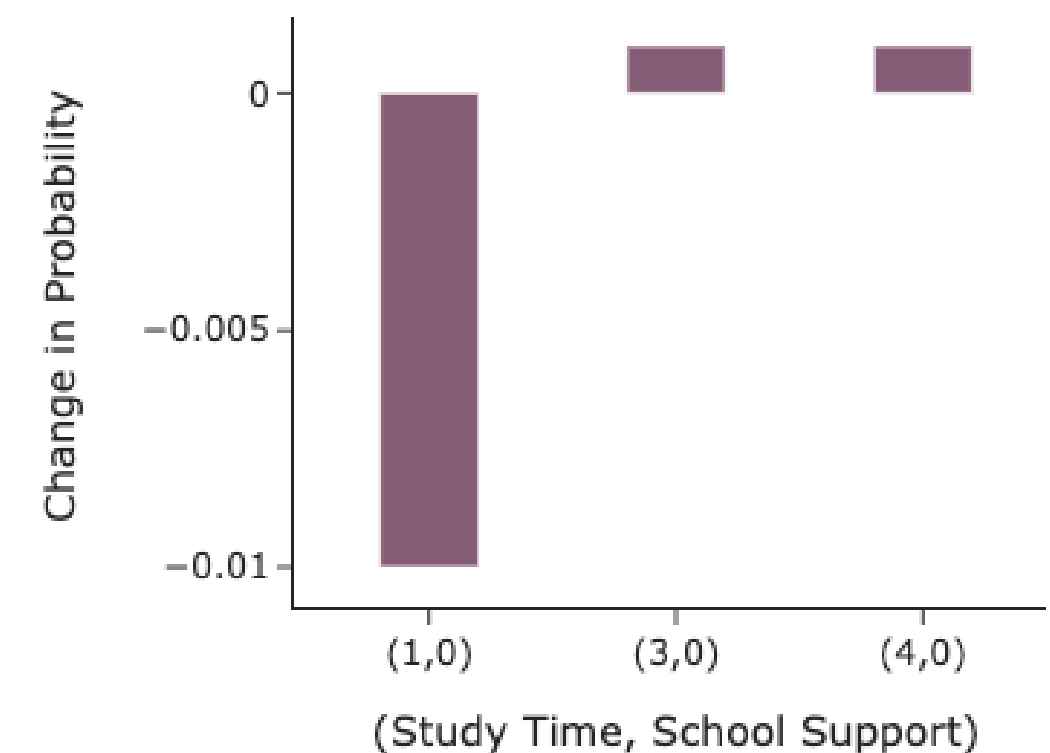
Initial probability of passing in G1 exam of Portuguese subject is 0.95

Initial study time value is 2 (studying between 2 & 5 hours per week)

Scenario with Study Time



Scenario with Study Time & School Support



Similarly, more scenarios were explored using multiple features for both the subjects

FINAL OUTCOMES



EXCEED CLIENT EXPECTATIONS

Client mentioned that
the team has achieved
all goals



REAL WORLD PROJECT EXPERIENCE



DEVELOPED MODEL FOR PREDICTING STUDENT'S GRADE



MAKES AN IMPACT

Developed Scenario
Modeling System to
provide suggestions to
students



Thank you!

**GROUP 25
STUDENT TEAM
THE UNIVERSITY OF MELBOURNE**
