

# What factors contribute to educational success?

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# Motivation



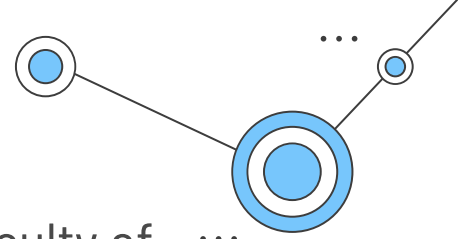
## Problem:

- 75% of high school students report that they ALWAYS or OFTEN worry about not getting good grades.
- However, there are numerous factors that can affect the end determination of our grades, some in our control and some out of it.
- In our project, we plan to explore this idea of whether we have full control over the grades we receive and how we can possibly predict our grades based on a series of factors.

## Motivating questions:

- Given these factors, how accurately can ML predict a student's grade?
- Which group of categories is most influential on a student's grade?  
(options: personal, educational, familial)

# Dataset Description



- A representative sample of **145 students** in 2019 in the Faculty of Engineering and Educational Sciences at the Near East Campus in Cyprus
- Students were given a questionnaire that consists of **30** questions that provide various info. about each student
- Each student also has a final grade of **0 - 7 (0 being an F, 7 being an A+)**

AGE GENDER HS\_TYPE SCHOLARSHIP WORK ACTIVITY PARTNER SALARY TRANSPORT LIVING ... PREP\_STUDY PREP\_EXAM

STUDENTID	AGE	GENDER	HS_TYPE	SCHOLARSHIP	WORK	ACTIVITY	PARTNER	SALARY	TRANSPORT	LIVING	...	PREP_STUDY	PREP_EXAM
STUDENT1	2	2	3	3	1	2	2	1	1	1	...	1	1
STUDENT2	2	2	3	3	1	2	2	1	1	1	...	1	1
STUDENT3	2	2	2	3	2	2	2	2	4	2	...	1	1
STUDENT4	1	1	1	3	1	2	1	2	1	2	...	1	2
STUDENT5	2	2	1	3	2	2	1	3	1	4	...	2	1

# Data Features

## Educational

high school type  
preparation - studied?  
preparation for exam  
takes notes  
listens  
likes to discuss  
likes flipped classroom  
cumulative gpa  
expected gpa  
weekly study hours  
reading frequency  
attendance to seminars  
attendance to classes  
preparation for midterm exams

## Personal

age  
gender  
scholarship  
work  
activity  
Partner  
Salary  
Transport  
Living  
regular artistic/sports activity  
total salary  
transportation to university  
accommodation type

## Familial

mother's education  
father's education  
number of brothers/sisters  
parental status  
(married/divorced)  
mother's occupation  
father's occupation

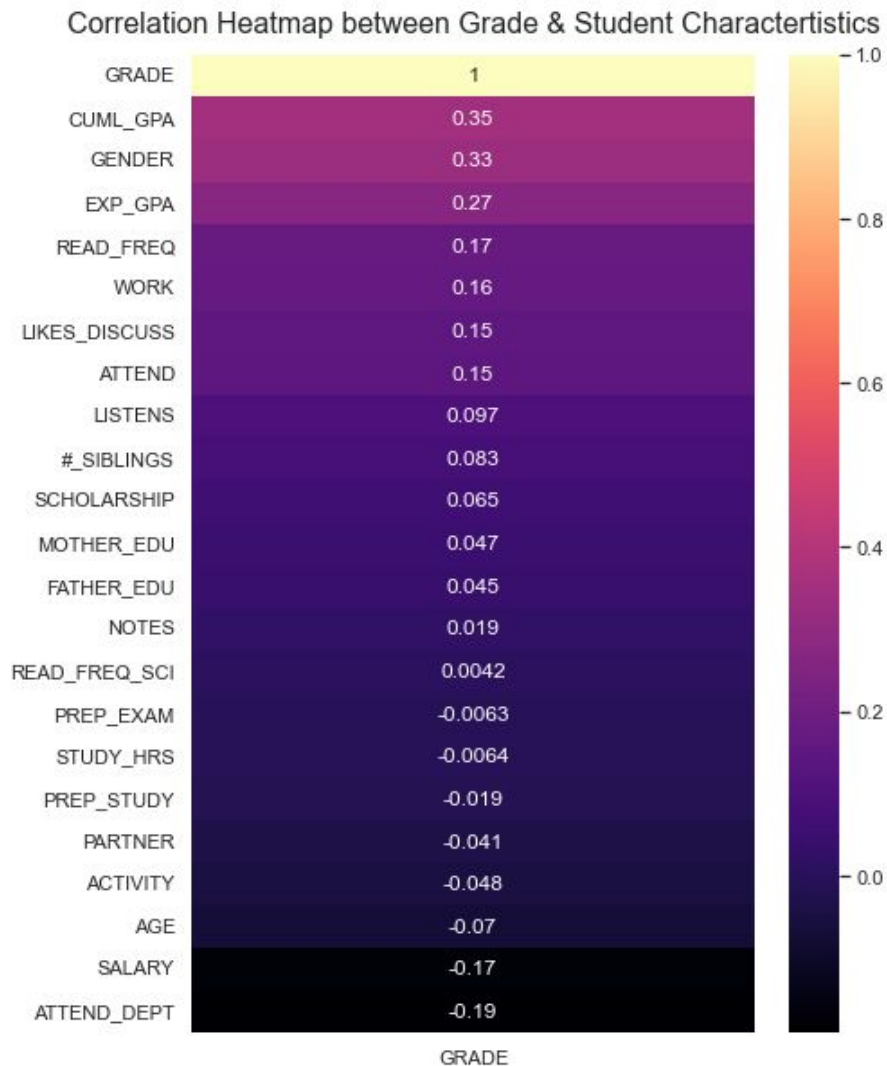


## Correlation Heatmap

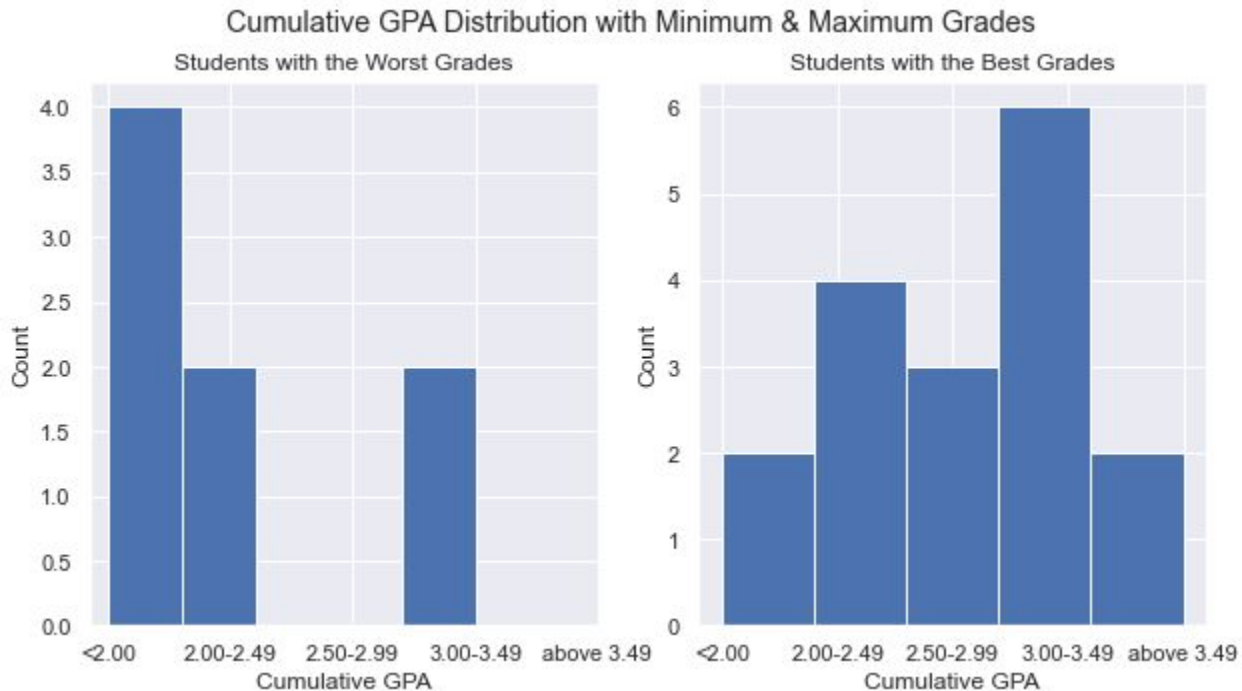
Applied to Student Characteristics  
with Ordinal or Binary Data

Highest POSITIVE correlation with  
grades : **Cumulative GPA (0.35)**

Highest NEGATIVE correlation with  
grades: **Attendance to Department  
Related Events (-0.19)**



# Cumulative GPA & Grades



# Machine Learning Methods

## k-NN Classifier

Cross Validated & K-Optimized

One for each category  
of features & One  
Overall

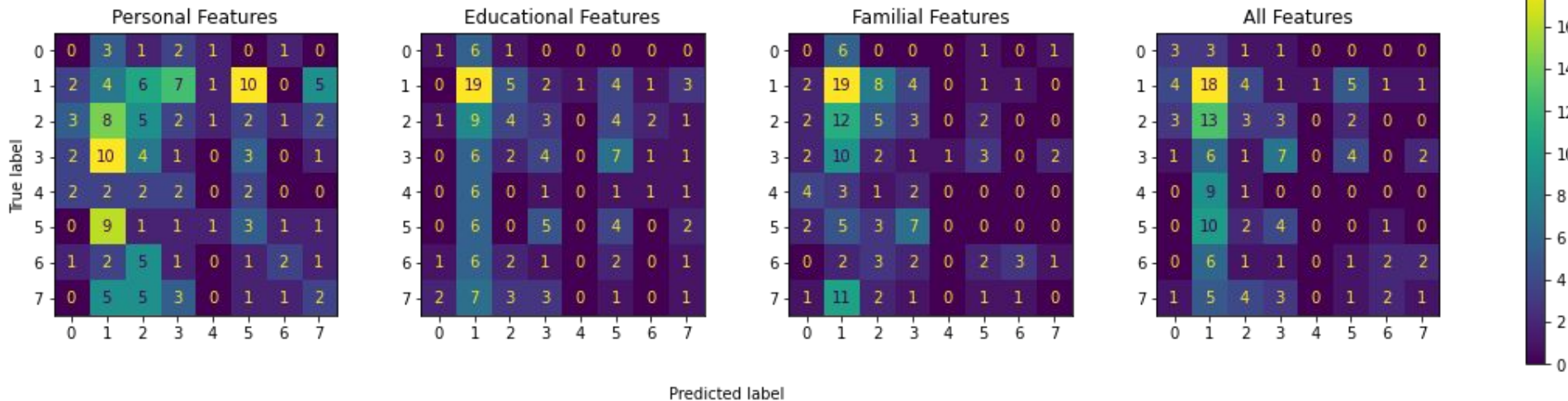
## Random Forest Regressor

Sample weighted

Fitted on ordinal and  
binary data

# Categorical Confusion Matrices

5-NN Classifiers



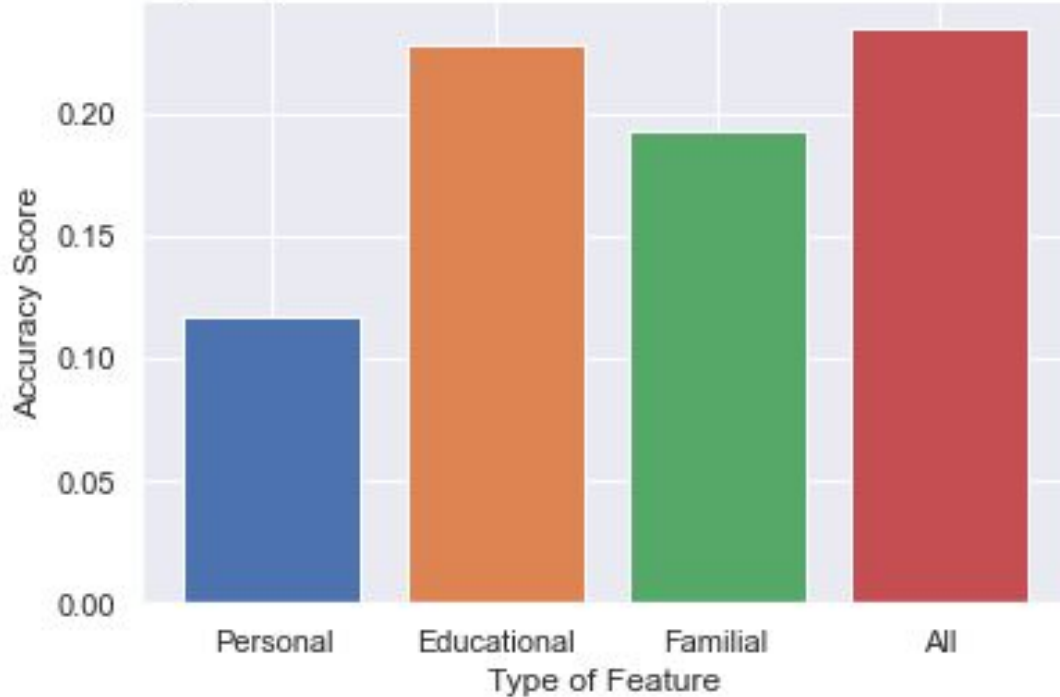




# Comparing Accuracy Scores



Accuracy Comparison Based on Feature Segmentation on 5-NN Classifier

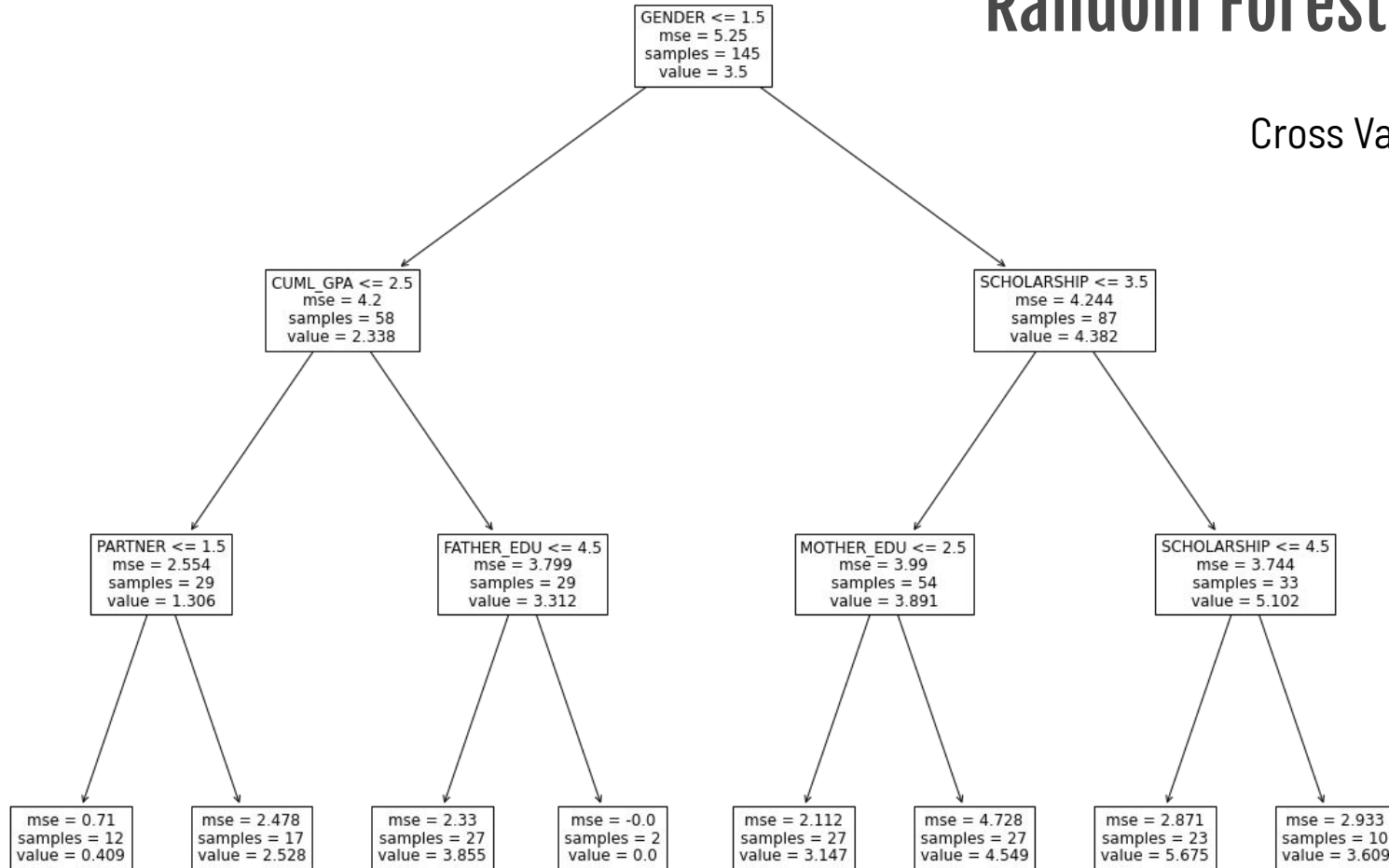


Accuracy Scores for Each Classifier:

- All Features: 0.23
- Personal : 0.11
- Familial: 0.19
- Education : 0.22

# Random Forest Regressor

Cross Validated  $R^2$ : .135





# Interpretation of Results



Random Forest Regressor:

- $R^2$  very low
  - ~13% of features (variables) helped predict grade
- No feature importance graph

k-NN Classifier:

- Low accuracy score throughout the matrices
- Education & Total Classifier had roughly equal accuracy scores
  - Education - most influential category



# Limitations

## **Random Forest Regressor:**

- Regressor would not output all grades
  - Our example tree only outputted grades from 0-5.

## **k-NN Classifier:**

- Could not properly weight samples based on sample size
  - Grade 1 was the most accurately predicted (highest sample size)
- Number of features could have played a role in helping Education have a higher accuracy score
  - More information for the classifier to work with
  - Had the Highest Number of Features

# Conclusions

Do not use this model

**Thanks for  
Watching!**

