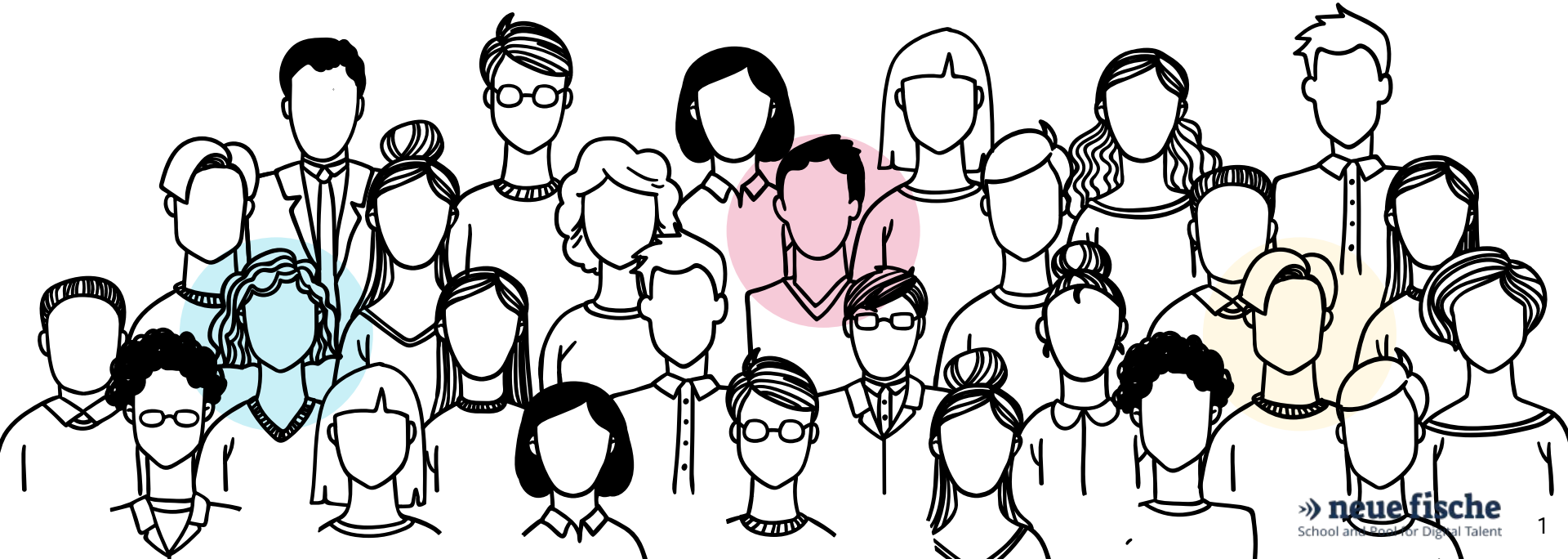
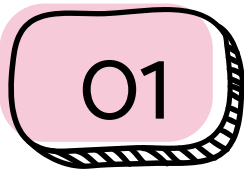


Personality and Drug Risk

By Matthis Westermann und Thomas Telgen





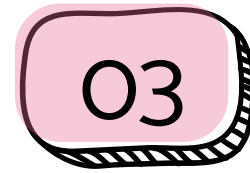
INTRODUCTION

What is it all about?



DATA SET

What are we working with?



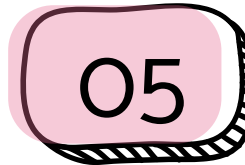
USER PREDICTION

Who uses cannabis?



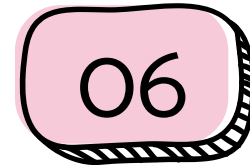
RESTRICTIONS

What did not work?



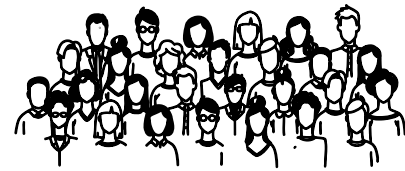
PERSONAL PREDICTION

What substances should you be careful with?



FUTURE WORK

What are the next steps?



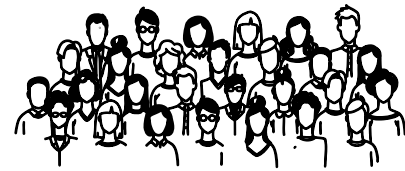
O1

INTRODUCTION

Correlation between personality and consumption habits?

Let's see what we can find in the [Drug Consumption Data Set!](#)





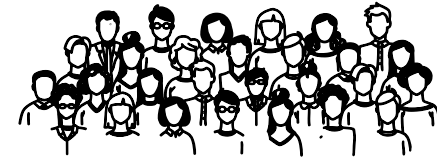
02

DATA SET

What are we working with?



SOME OVERVIEW



1885 Participants

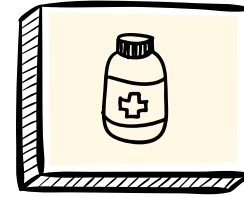
32 Features including:

- Age
- Education Level
- 7 Personal Traits
- Use frequency for 18 Drugs
- One made up Drug



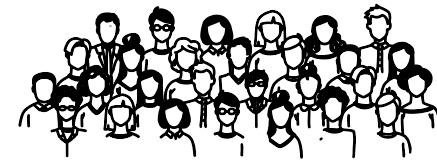
Gender

Female	942
Male	943

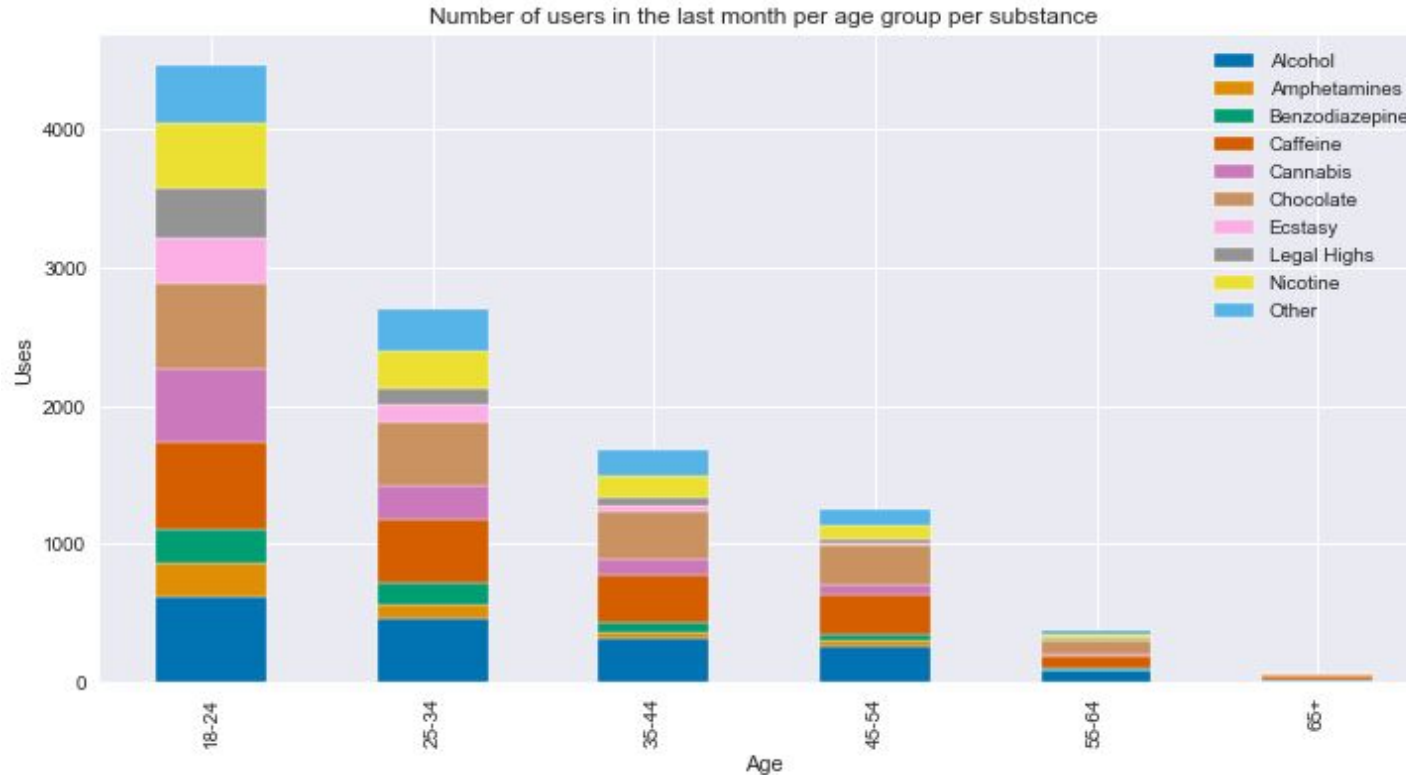


Source

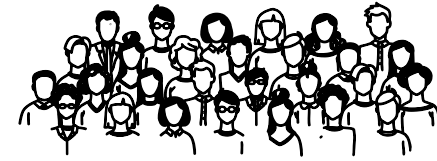
- Online Survey
- Mostly from UK and USA
- From 3/2011 to 3/2012



USE FREQUENCY BY AGE

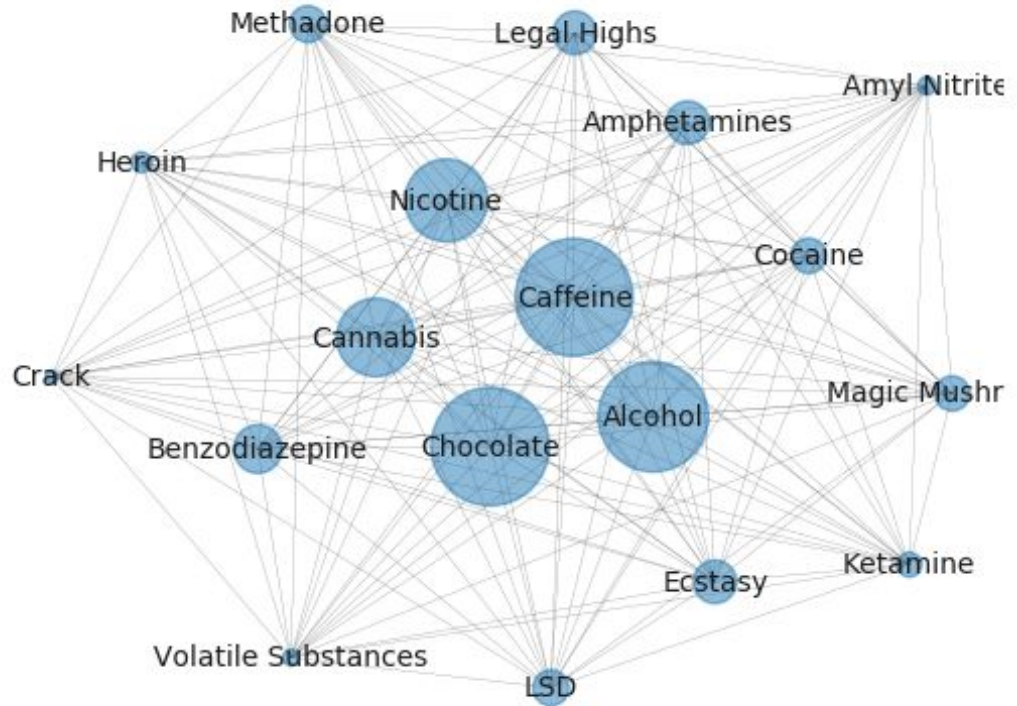


SUBSTANCE FRIENDS?



Findings:

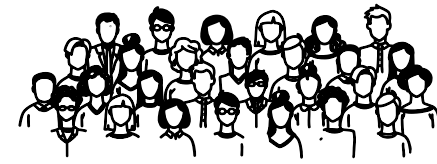
- Four legal substances tend to cluster with Cannabis
- Other patterns not stable
- Cannabis and Ecstasy are the most linked illegal substances



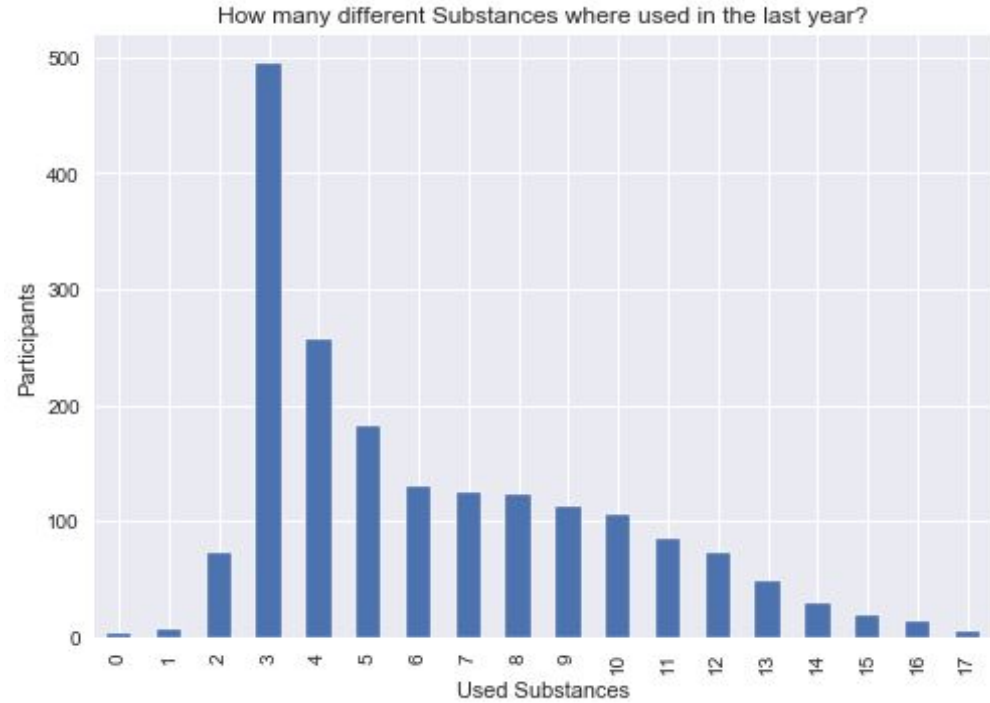
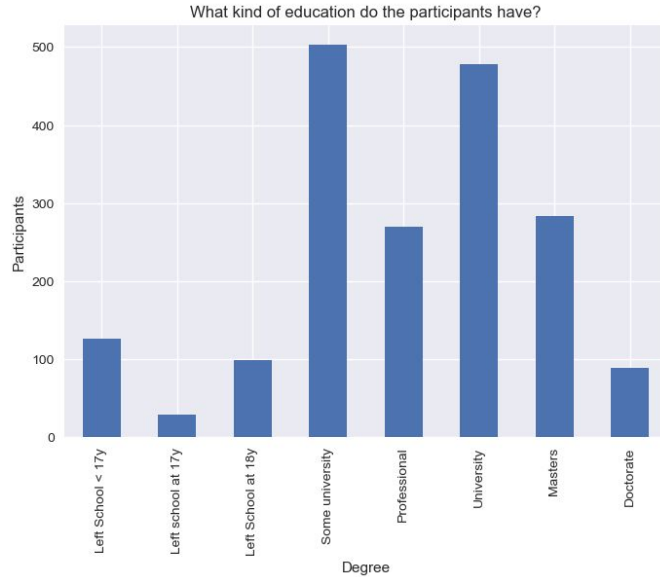
Explanation of graph:

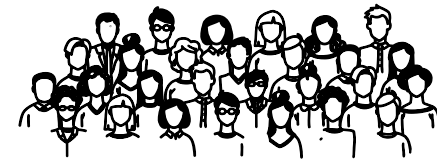
More commonly used substances have larger nodes.

Substances more regularly used together are closer to each other.



WHAT ELSE?



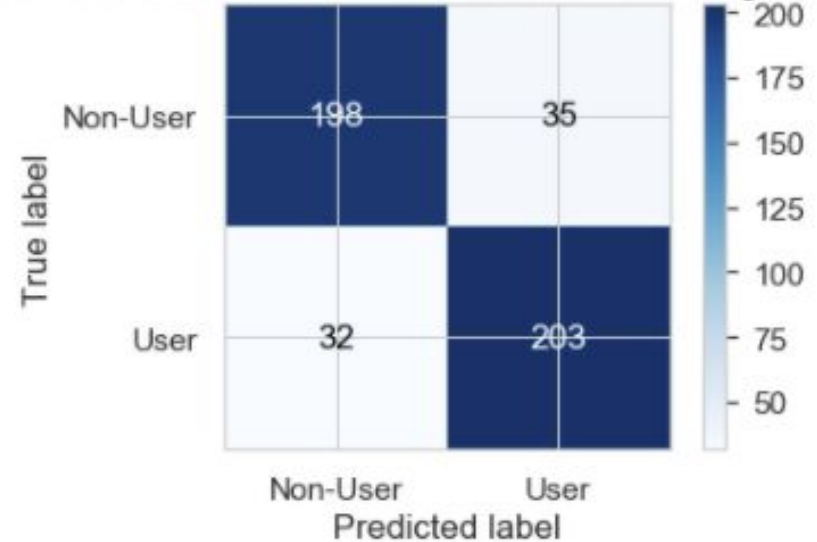


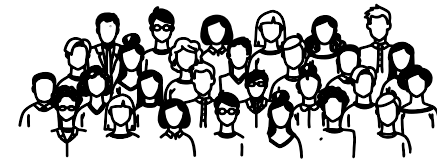
03 - SUBSTANCE USER PREDICTION

- Drug use in binary classes (uses drugs, doesn't use drugs)
- Many imbalanced datasets which makes it hard to predict these values

Best models → SVC, Adaboost, Gradient Boosting

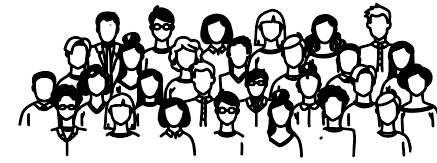
Cannabis user confusion matrix: Gradient Boosting Classifier





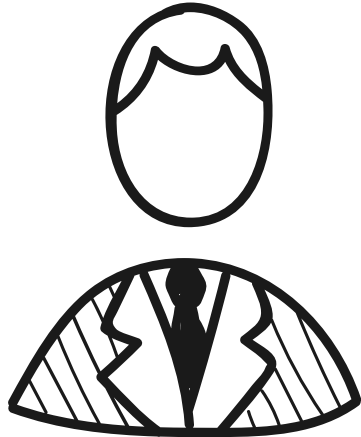
O4 - PREDICTIVE MODELING - RESTRICTIONS

- Drug use in multiple classes (ordinal values from 0 to 6)
- There were no predictions which were really good
 - Legal drugs → caffeine
 - Soft drugs → amyl nitrite, lsd
 - Hard drugs → cocaine, meth
- Amyl nitrite, lsd and vsa have a lighter imbalance than alcohol or crack
- Can be predicted better by multiclass classification

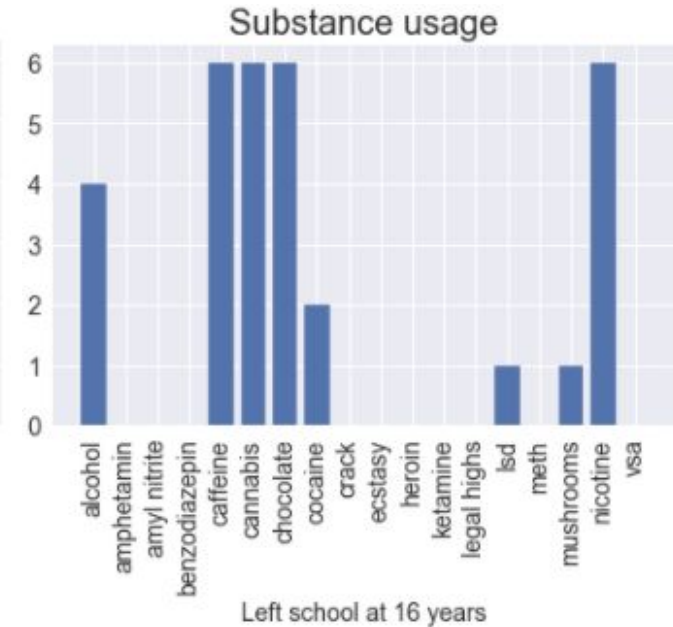
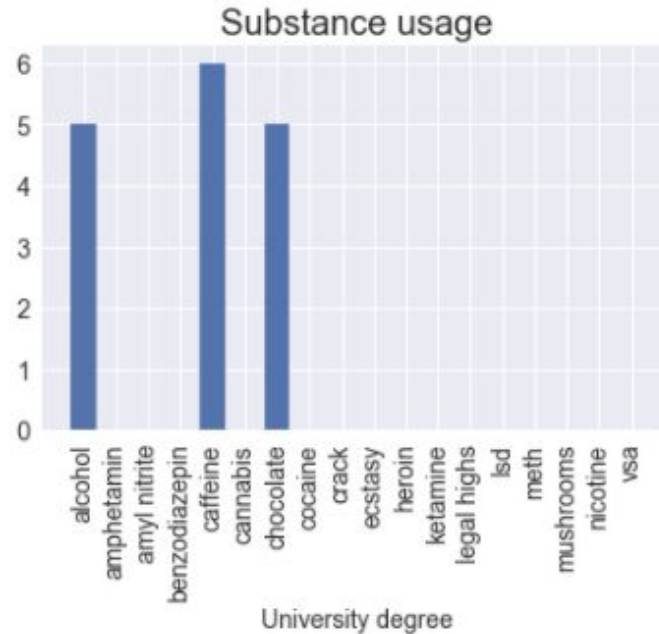


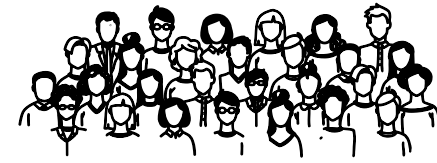
O5 - WHAT SUBSTANCES FOR ME?

- Personal data
- Changing the educational background
- Model KNN



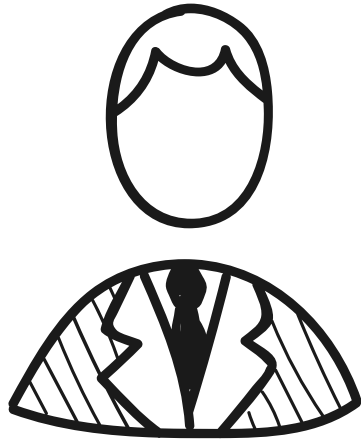
Test Subject A



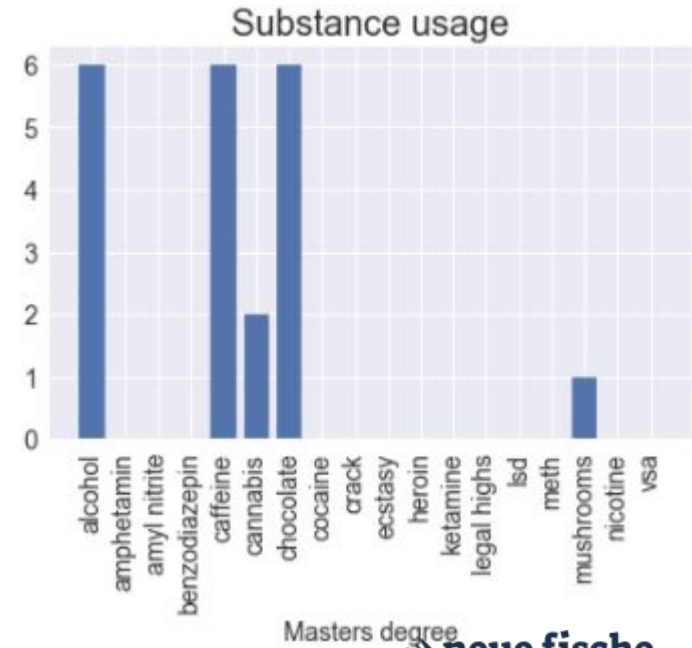
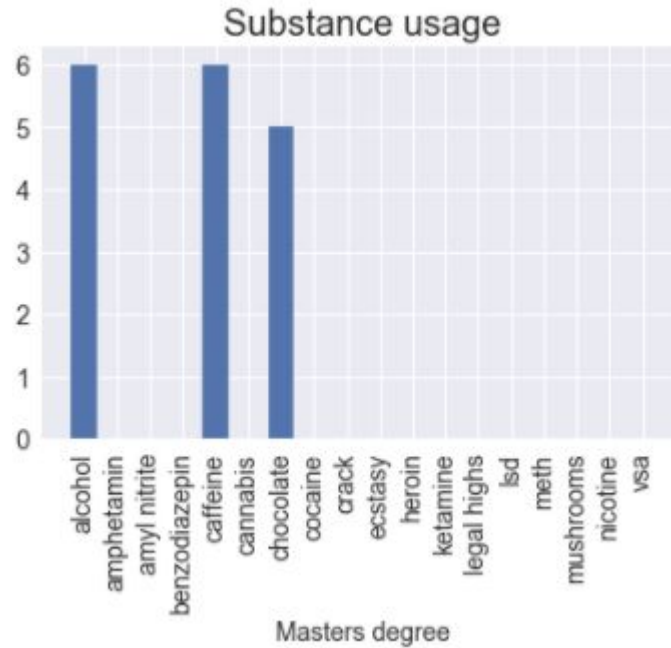


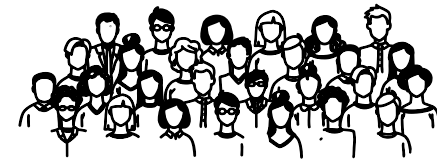
AND FOR HIM?

- Personal data
- Changing to higher impulsiveness and higher sensation seeking score
- Model KNN



Test Subject B





FUTURE WORK

What we'd like to try:

- Improving prediction by favoring substances with less occurrences (SMOTE)
- Focus on feature importance
- Developing an easy to use implementation for the substance “recommendation”

Questions to answer:

- What kind of substances are becoming less interesting over time?