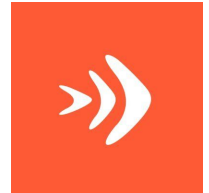




Interim Report: **Detecting Drug Consumption Risk**

A machine learning project by Mirko Knoche and
Nina Notman, by request of the *American
Psychology Association (APA)*



REQUEST BY THE APA



REQUEST BY THE APA

“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a future time point.



REQUEST BY THE APA

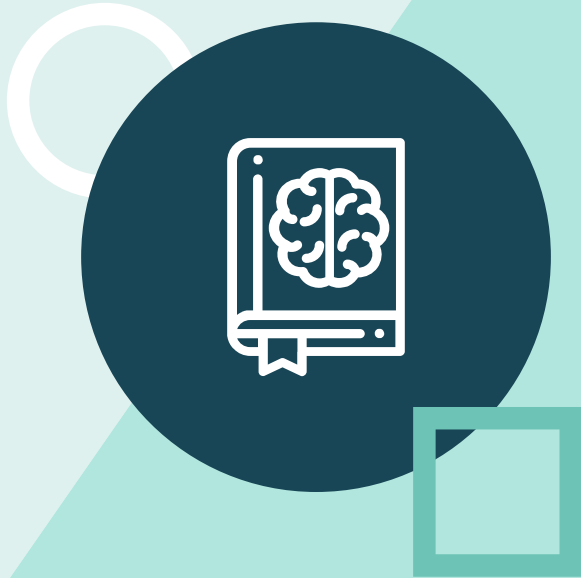
DATASET BY ELAINE
FEHRMAN

“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a future time point.



THE DATASET



FEATURES

DEMOGRAPHICS

Age, Gender,
Education, Ethnicity,
Country

NEO-FF-R

Neuroticism,
Extraversion,
Openness,
Agreeableness,
Conscientiousness

BIS-11 & ImpSS

Impulsiveness,
Impulsiveness-
Sensation-Seeking

LABELS

PSYCHOTIC DRUGS

Alcohol, Amphetamines, Amyl Nitrite, Benzodiazepines, Cannabis,
Chocolate, Cocaine, Caffeine, Crack, Ecstasy, Heroin, Ketamine,
Legal Highs, LSD, Methadone, Mushrooms, Nicotine, VSA

PREDICTION OF CANNABIS CONSUMPTION RISK

“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a future time point.



PREDICTION OF CANNABIS CONSUMPTION RISK



MODEL SELECTION

- K-nearest Neighbors
- Logistic Regression
- Random Forest Classifier
- Support Vector Classifier
- XGBoost

PREDICTION OF CANNABIS CONSUMPTION RISK



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MODEL EVALUATION

- Classification Report
- Confusion Matrix
- ROC Curve & AUC Score

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MODEL EVALUATION

- Classification Report
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- ROC Curve & AUC Score



MODEL TUNING

- Grid Search (scorer = F1)
- Feature Importance

“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a future time point.



“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a feature time point.

BEST MODEL

XGBoost after GridSearch



“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a feature time point.

BEST MODEL

XGBoost after GridSearch



“

I wish I could identify patients at risk for cannabis consumption so it won't influence therapy at a feature time point.

BEST MODEL

XGBoost after GridSearch

- Identify 4 out of 5 correctly as cannabis consumer.
- Fine. But let's not leave the one hanging. We can do better!



PREDICTION OF **HIGHLY ADDICTIVE** DRUG CONSUMPTION RISK

PREDICTION OF HIGHLY ADDICTIVE DRUG CONSUMPTION RISK

CHANGE
LABEL

HIGHLY ADDICTIVE

PSYCHOTIC DRUGS

Ecstasy, Heroin, LSD, Meth, Cocaine

PREDICTION OF HIGHLY ADDICTIVE DRUG CONSUMPTION RISK

CHANGE
LABEL

HIGHLY ADDICTIVE
PSYCHOTIC DRUGS

Ecstasy, Heroin, LSD, Meth, Cocaine

SAME
FUNCTION



MODEL SELECTION

- K-nearest Neighbors
- Logistic Regression
- Random Forest Classifier
- Support Vector Classifier
- XGBoost



MODEL EVALUATION

- Classification Report
- Confusion Matrix
- ROC Curve & AUC Score



MODEL TUNING

- Grid Search (scorer = F1)
- Feature Importance



“

Request:

Identify patients that are likely to take highly addictive drugs and thus won't be able to focus on therapy...

BEST MODEL

SVM after GridSearch

Only 75% accuracy on predicting consumption of highly addictive drugs.

Future Work: Feature Engineering

WATCH OUT!

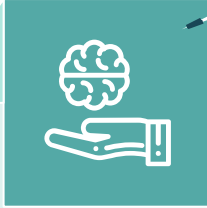
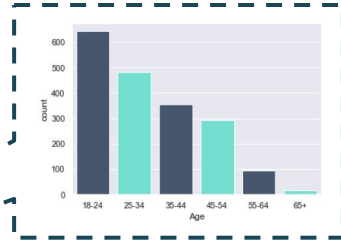


WATCH OUT!

ONLY FOR **WHITE**, **ENGLISH** SPEAKING PEOPLE!



FUTURE WORK

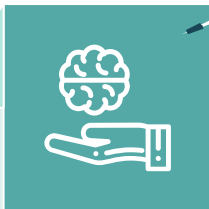
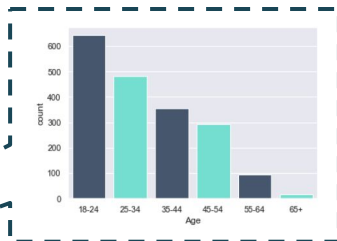


DISTRIBUTION

Age / Country / Ethnicity



FUTURE WORK



DISTRIBUTION

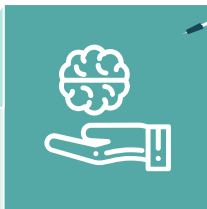
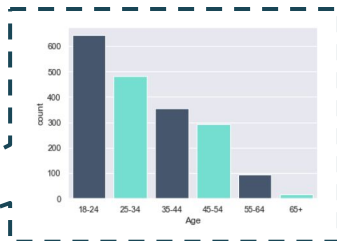
Age / Country / Ethnicity



COLLECT MORE DATA

Expand data collection to
optimize and generalize model

FUTURE WORK



DISTRIBUTION

Age / Country / Ethnicity



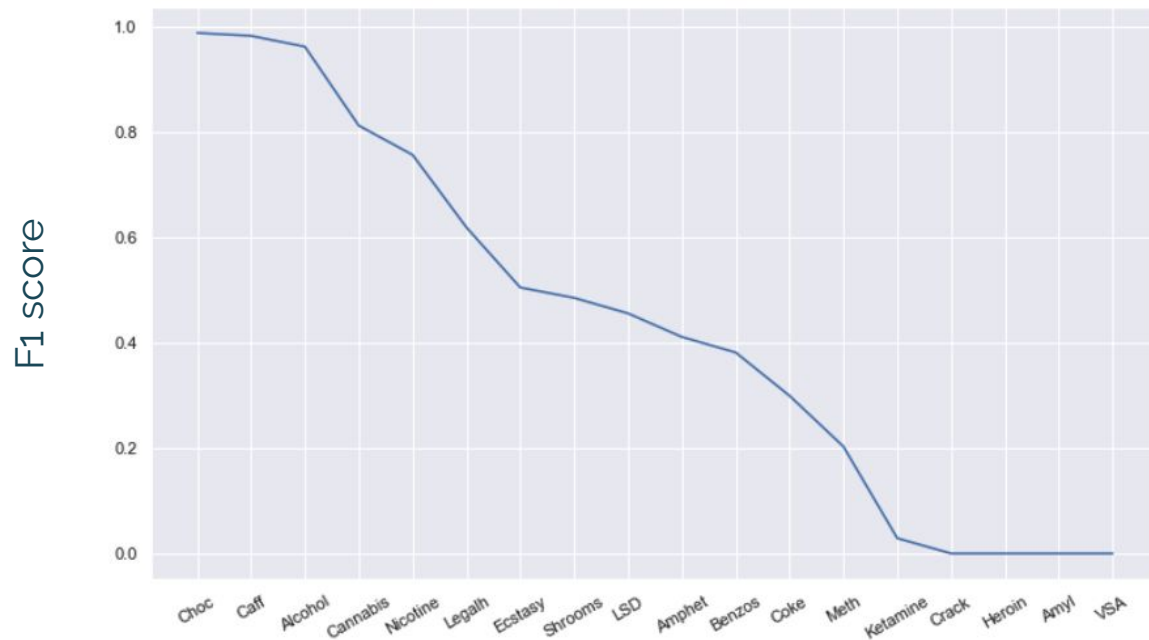
COLLECT MORE DATA

Expand data collection to
optimize and generalize model

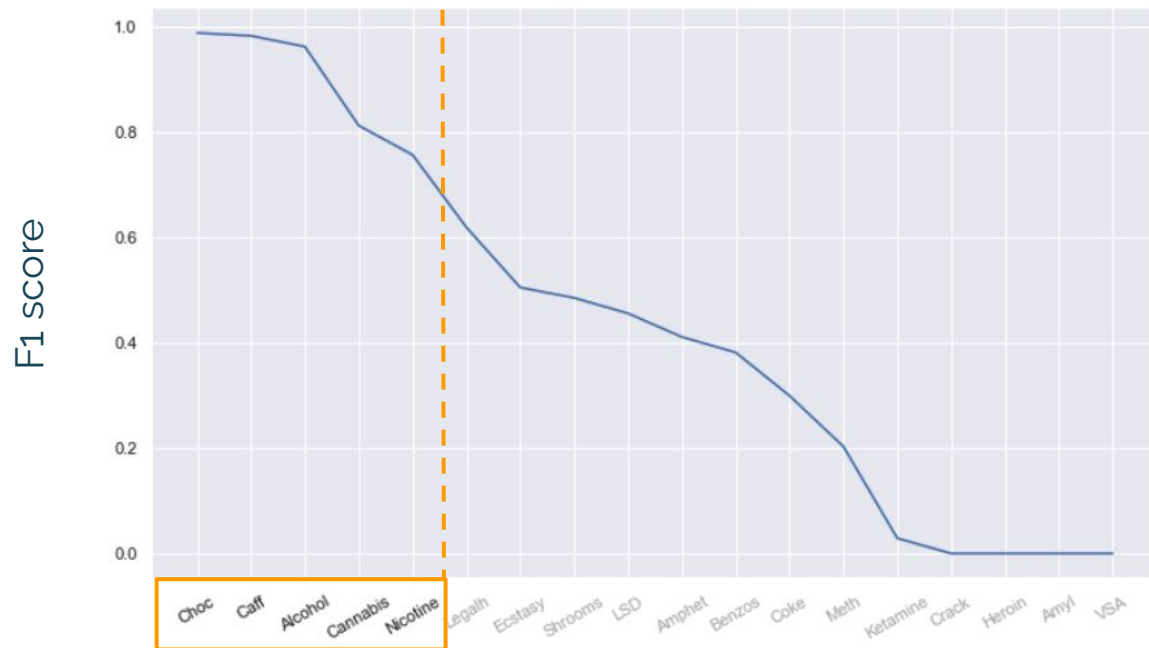


FEATURE ENGINEERING

GROUPING OF DRUGS



GROUPING OF DRUGS





LET'S TALK!



Nina Notman

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Mirko Knoche

mirko.knoche@posteo.de

THANK YOU!

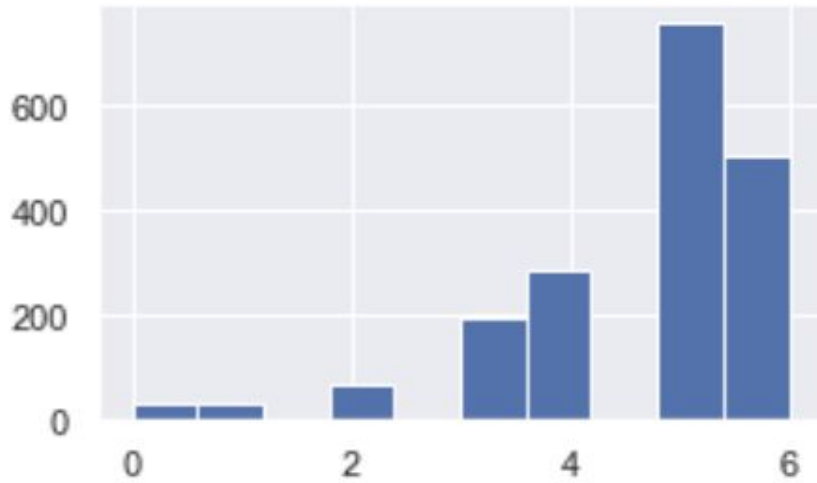
Questions?



Appendix

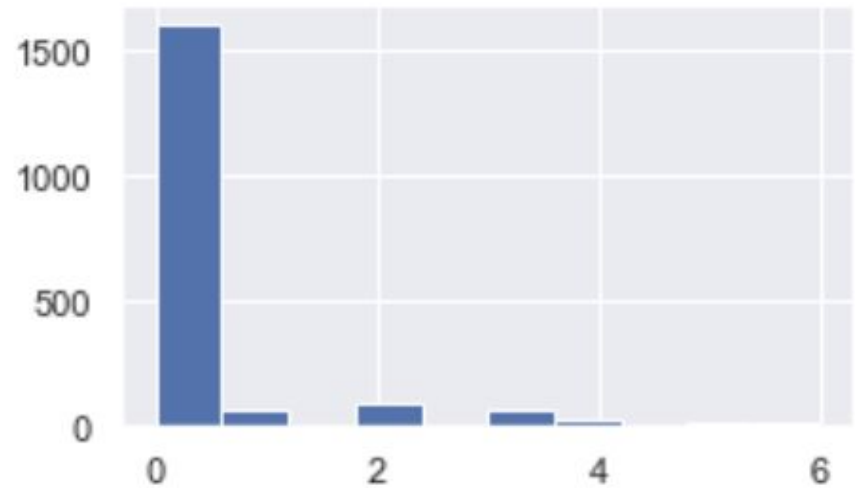
Distribution of drug use

Alcohol



High use

Heroin



Low use