

# How can Personal Data Predict Drug Use?

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#### Introduction

In this project, we have a dataset of 1885 data points categorised into a mix of 34 continuous & categorical measurements and drug use readings. Some notable measurements include

- categorical: age group, ethnicity, UseLevel ...
- numerical: opentoexperience, impulsiveness

In this poster, we will see exploratory analysis of the data, a K-Nearest-Neigbours classification for Use-Level and a random forest to predict cannibis use.

Credit and thanks for the poster template goes to Emma McCoy.

# **Exploratory Analysis**

Below are some exploratory plots comparing different measurements



## Predicting Cannabis use by Drug Use & Personality Metrics

Pose the following questions:

- Q1: what information do we need to know about someone to predict their propensity to use cannabis dictors regularly?

  Method
- Q2: Does one's country affect the likelihood one uses cannabis regularly?

Method: Train a random forest to sort for best predictors - is country significant?

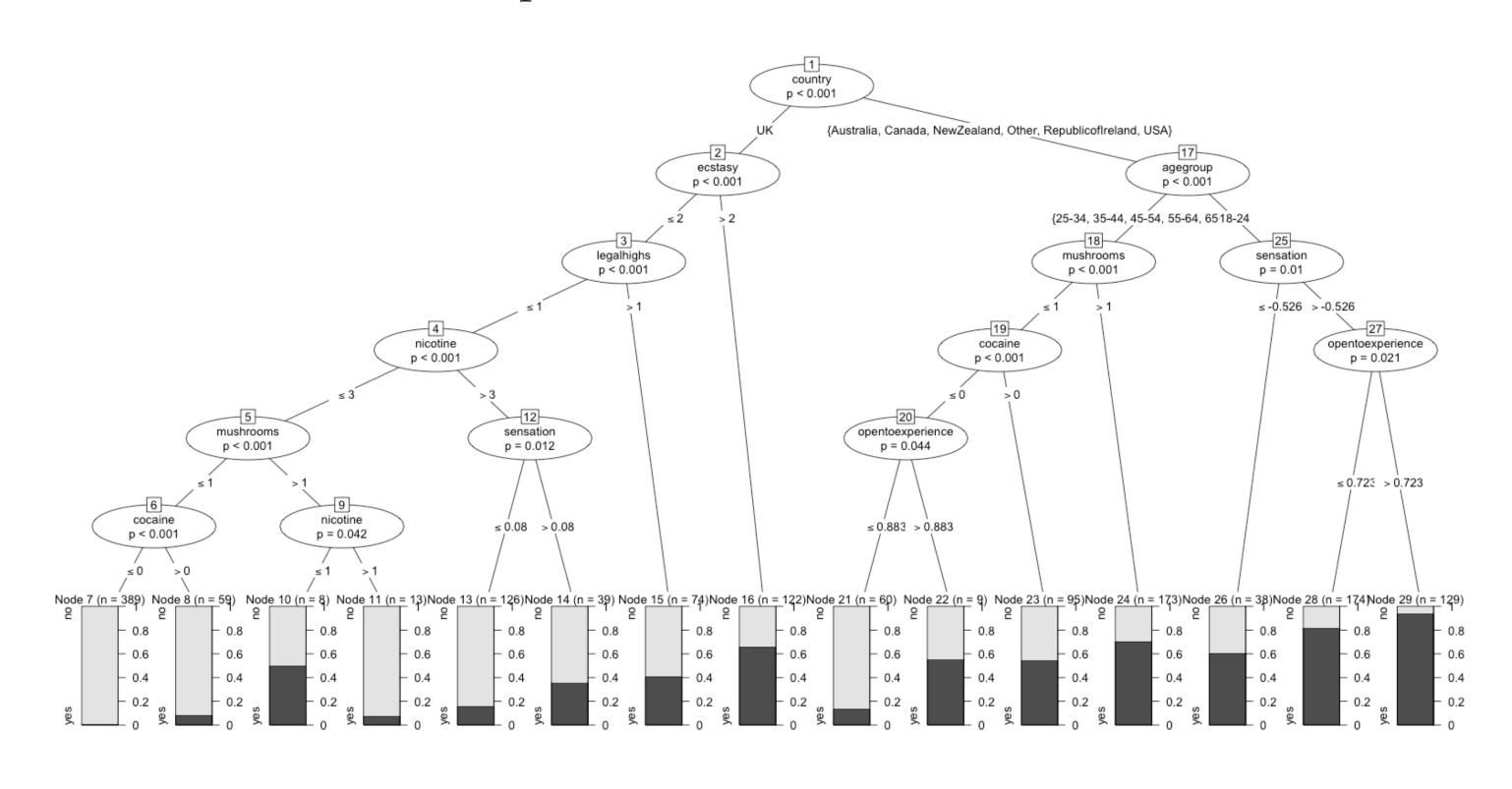
- Data in: 16 personality predictors, drug information
- Response: "yes" regular cannabis use; "no" otherwise

RFcan <- randomForest(cannabis . , data=can.train, importance = TRUE, ntree = 2000)

can.prediction <- predict(RFcan, can.test)</pre>

## Example Tree

Here we see the visualisation of an example tree



- Method ran once
- Importance function called
- Method ran again for 10 most significant predictors

#### Results

	MeanDecreaseAccuracy	
agegroup	58.37245	
gender	26.39574	
country	104.30951	
opentoexperience	32.24647	
sensation	47.30529	
nicotine	53.36737	
cocaine	39.95916	
ecstasy	61.13762	
legalhighs	58.79608	
mushrooms	69.93532	

Bigger MDA means larger significance to random forest prediction

#### Answers

First run: Accuracy = 0.8381 Second run: Accuracy = 0.8170

Conclusion: Model runs well for sparser predictions

**A1**: The 10 in the results section are most likely a good starting point

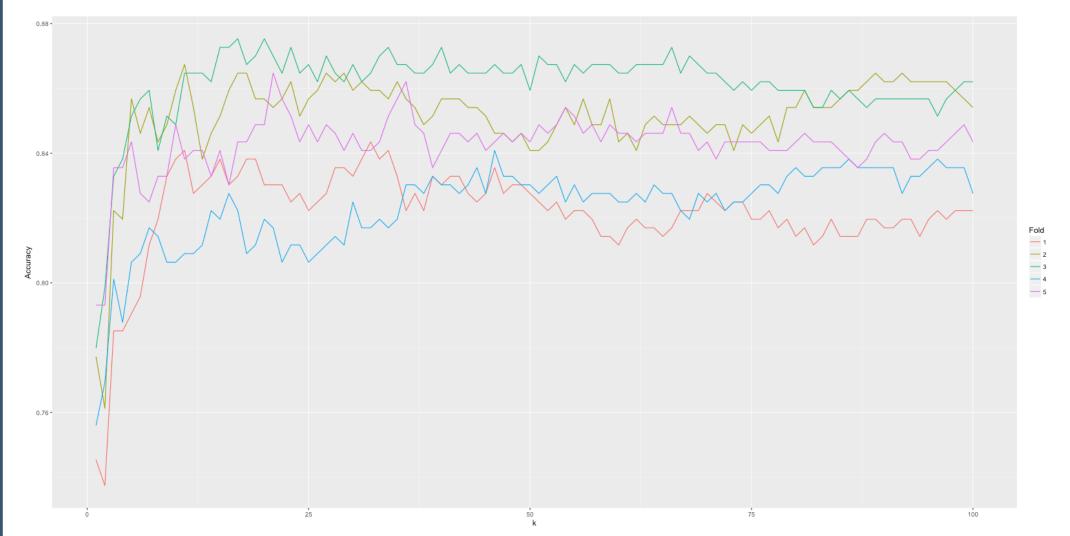
**A2**: One's country appears to be a very significant predictor for regular cannabis use

### Using KNN to Predict UseLevel

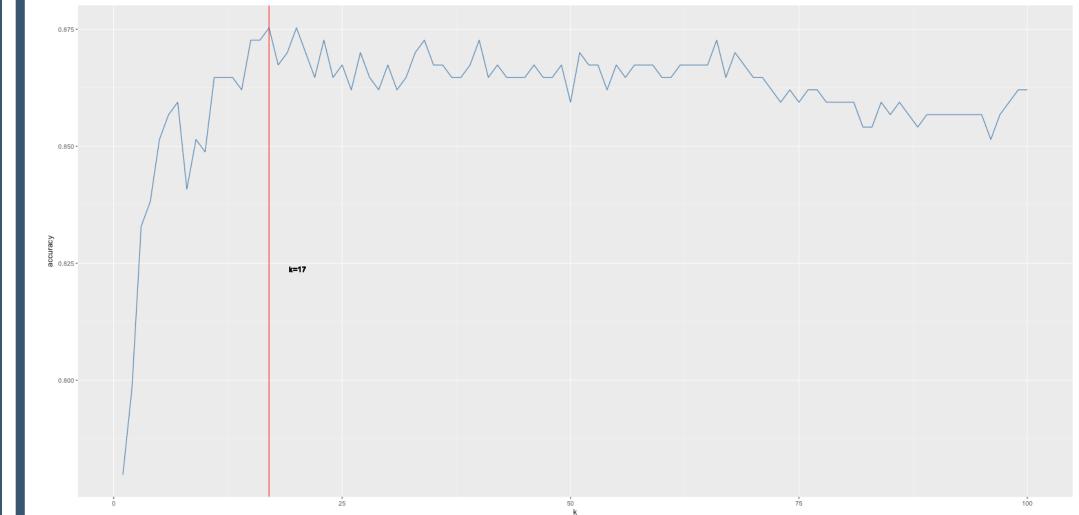
**Q**: Predict UseLevel using first 16 personality predictors

**Method**: Use KNN. Perform 5-cross-fold-analysis and vary 'k' parameter at the same time. Pick best fold & best k to predict

predictions <- knn(train.nk[, -1],
test.nk[, -1], train.nk\$UseLevel, k =
k)</pre>



- Pick best k and best fold



**Results**: Accuracy = 0.8753

prediction	low	high
low	151	23
high	24	179

## Significant Predictors

Exploratory results and linear regression show the following are significant predictors for drug use

- Gender
- Opentoexperience
- Sensation
- Nicotine
- Country
- agegroup (potentially)