

# Design of a diagnosis and follow-up platform for patients with chronic headaches

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Counsellor: ing. Olivier Janssens

Faculty of Engineering and Architecture

Intro

Current process Ghent  
University Hospital

Platform requirements

Mobile application

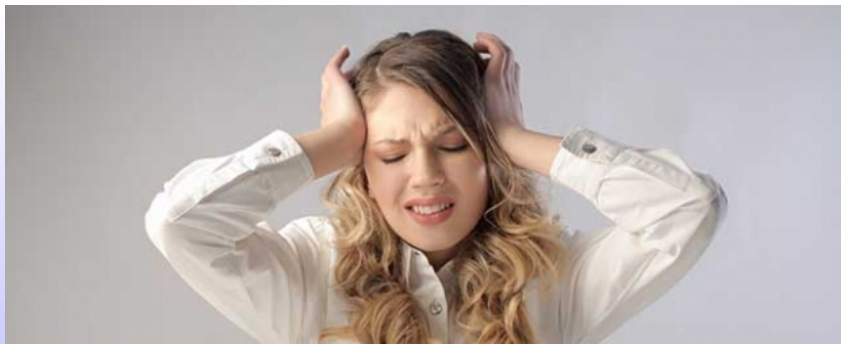
Backend and data exposure

Machine learning

Doctor dashboard

Conclusion & future work

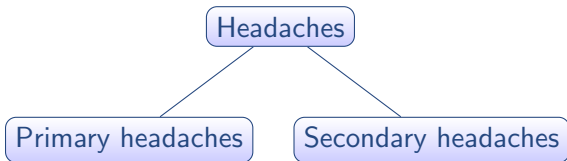
# Headaches



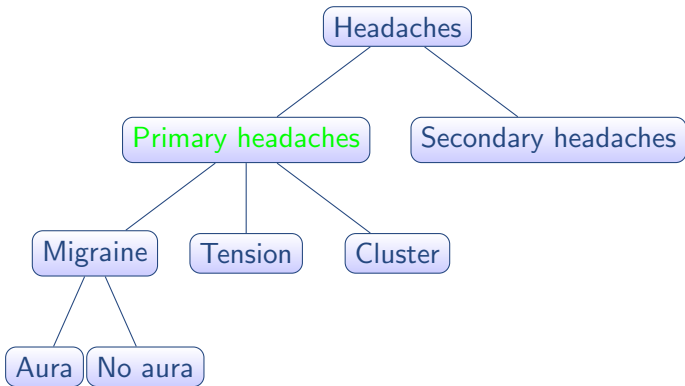
# Headaches

Headaches

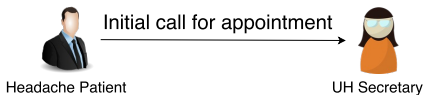
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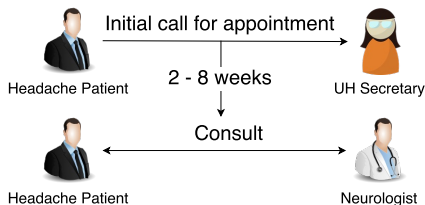
# Headaches



# Current process at Ghent UH

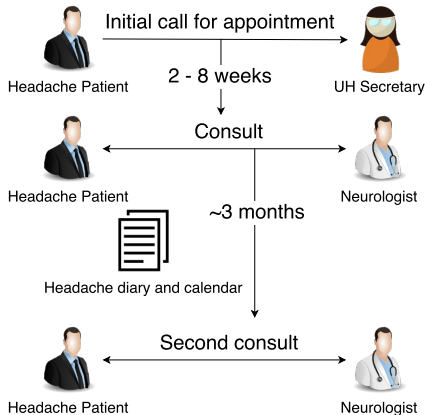


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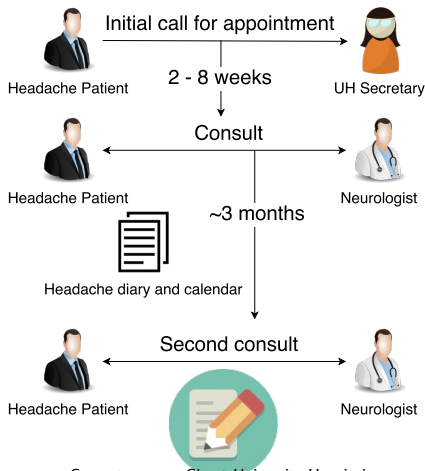




## Current process at Ghent UH



# Current process at Ghent UH



# Current process Ghent University Hospital

Current process at Ghent University Hospital is:

- ▶ Not digital
- ▶ cumbersome
- ▶ time consuming

So there is need for a better (digital) alternative! This alternative has to:

- ▶ capture at least the same information as current solution
- ▶ be more efficient
- ▶ support doctors in forming a diagnosis

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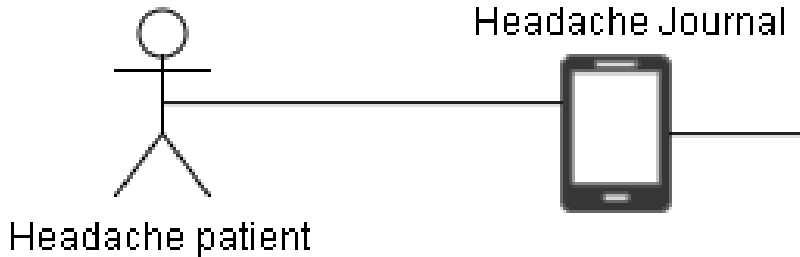
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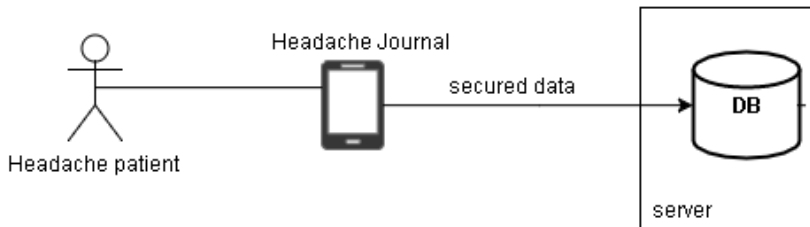
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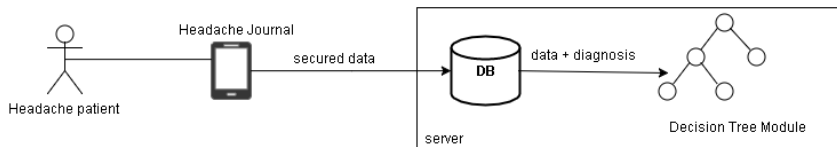
## Platform requirements



# Platform requirements

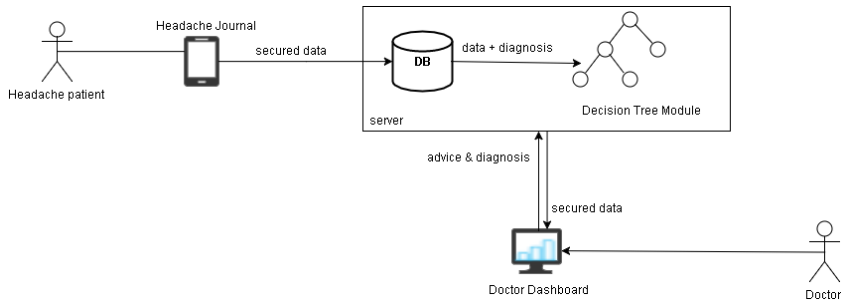


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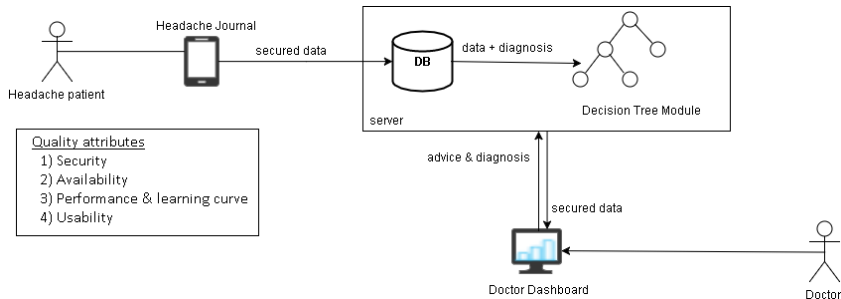




# Platform requirements



# Platform requirements



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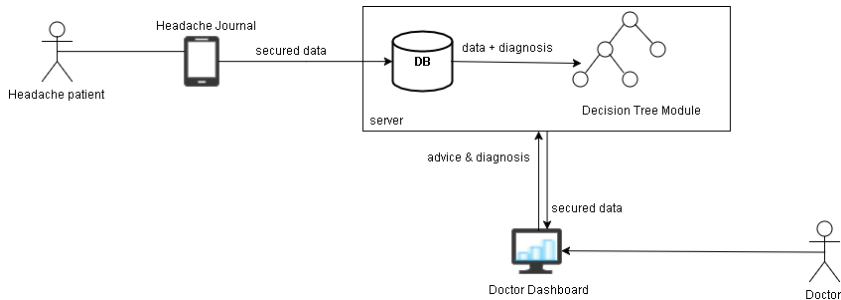
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Mobile application  
Chronicals

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# Mobile Application



# Mobile Application



# Mobile Application

Why create a new application?

## Competition

- ▶ Migraine Buddy
- ▶ Headache Diary
- ▶ Pfizer headache journal

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All good, but:

# Mobile Application

Why create a new application?

## Competition

- ▶ Migraine Buddy
- ▶ Headache Diary
- ▶ Pfizer headache journal

All good, but:

- ▶ none captures all data needed
- ▶ none offers usable data export

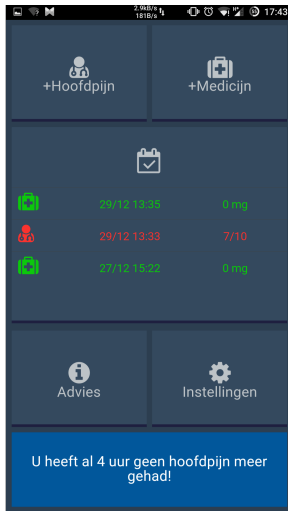


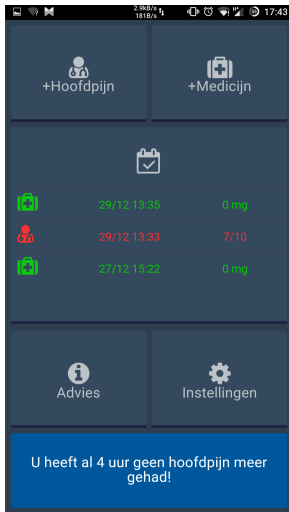
## Cross platform vs Native

	Native	Cross-platform
+	+ Native UX	+ 1 language
	+ device-specific features	+ Write once, run everywhere
	+ Better performance	+ Less maintenance
-	- Multiple languages	- Slower (lower performance)
	- Time consuming (development)	- Less device specific features
		- Harder to release online (Play Store/App Store)

## Cross platform vs Native

	Native	Cross-platform
+	<ul style="list-style-type: none"> <li>+ Native UX</li> <li>+ device-specific features</li> <li>+ Better performance</li> </ul>	<ul style="list-style-type: none"> <li>+ 1 language</li> <li>+ Write once, run everywhere</li> <li>+ Less maintenance</li> </ul>
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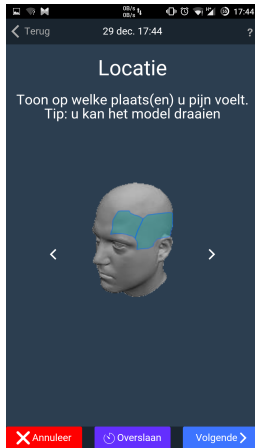
# Chronicals



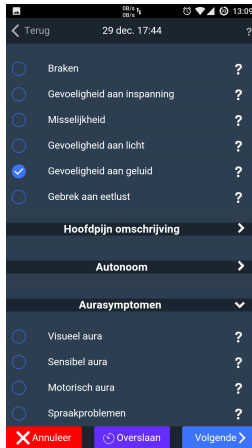
# Chronicals



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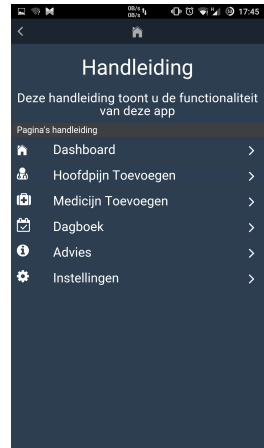
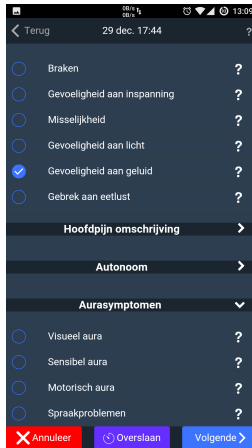
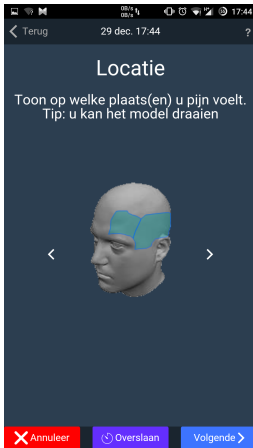


Mobile application



Chronicals

# Chronicals



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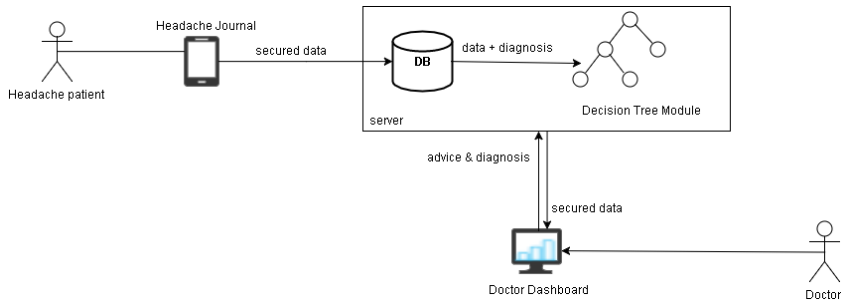
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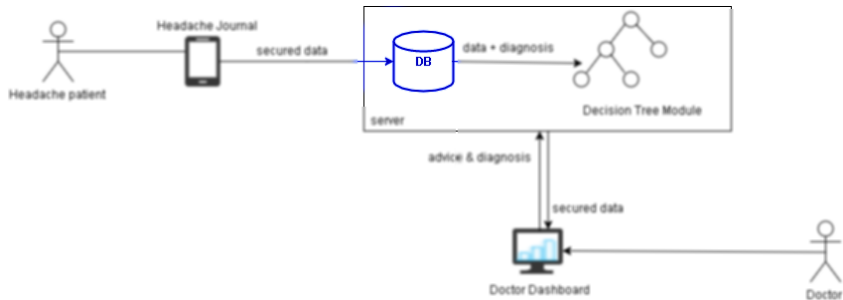
Backend and data exposure



# Backend and data exposure



## Backend and data exposure

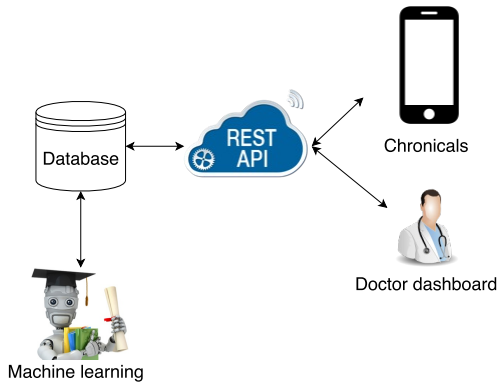


# Backend and data exposure

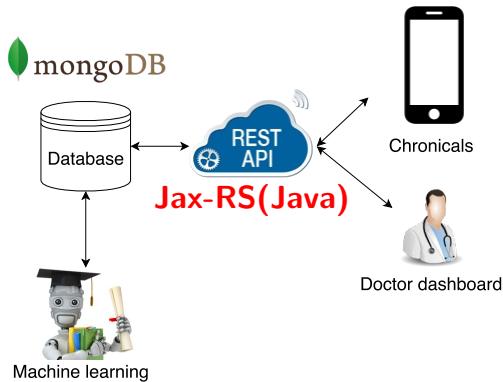
## Components

- ▶ Database
- ▶ Connection to App
- ▶ Connection to Doctor Dashboard
- ▶ Connection Machine learning module

# System



# System



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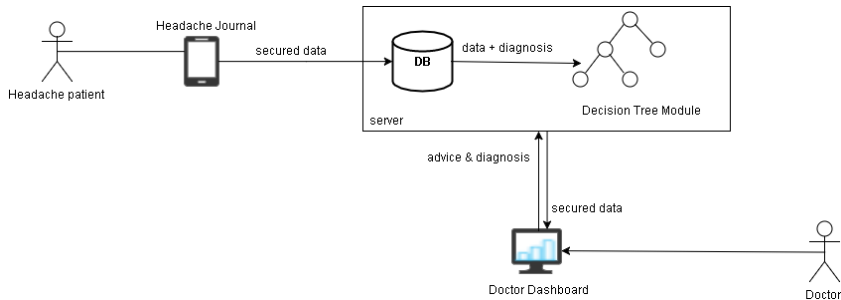
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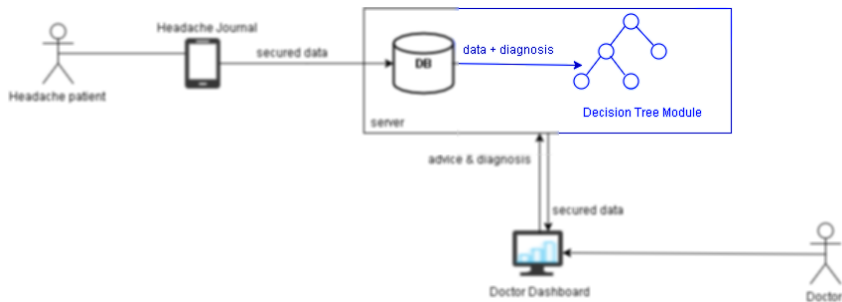
Conclusion & future work

Backend and data exposure

# Machine learning



# Machine learning





# Machine Learning

Decision support ( $\neq$  decision making)  $\Rightarrow$  White box model

## Possible models

- ▶ Decision trees
- ▶ Random Forests (Gray box)
- ▶ Bayesian networks

# Machine Learning

Decision support ( $\neq$  decision making)  $\Rightarrow$  White box model

## Possible models

- ▶ Decision trees
- ▶ Random Forests (Gray box)
- ▶ Bayesian networks

# Many different DT induction algorithms



C4.5 (C5.0)



CART



QUEST

...

→ **Which tree is the most beautiful?**

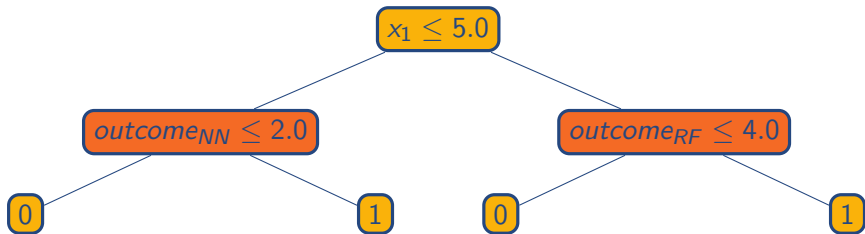
# Current ensembles lack interpretability

**Boosting, bagging, random forests, etc.** require majority voting (classification) or mean calculation (regression) to obtain prediction



## Current ensembles lack interpretability

The final decision tree obtained by **stacking** contains uninterpretable internal nodes



# An ensemble technique WITH interpretability

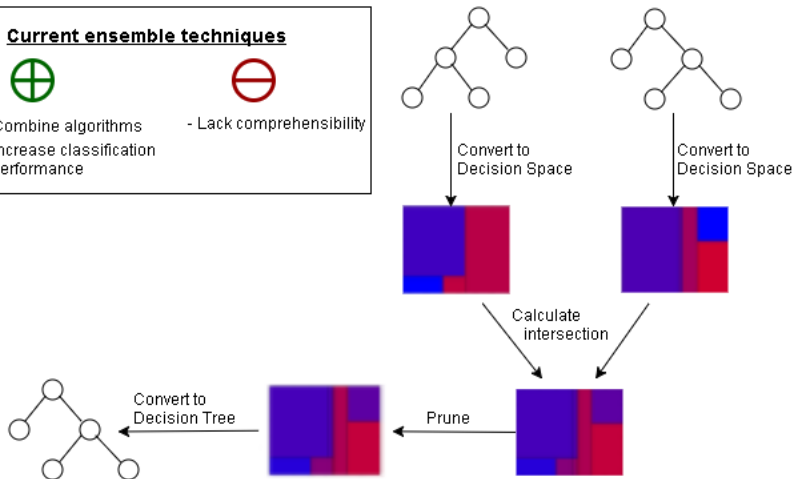
## Current ensemble techniques



- Combine algorithms
- Increase classification performance



- Lack comprehensibility

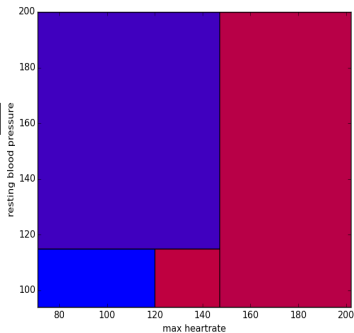
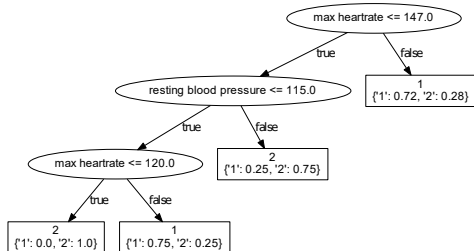


## Decision tree $\rightarrow$ decision space

### Converting decision trees to decision spaces

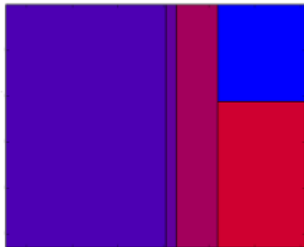
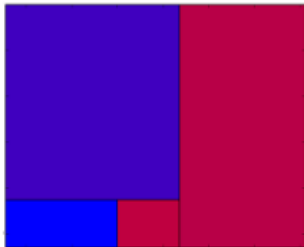
We can define a one-to-one mapping between a decision tree and a set of  $k$ -dimensional hyperplanes ( $k = \# \text{features}$ ), called **decision space**. Each node in the decision tree corresponds to a hyperplane in the decision space.

## Decision tree → decision space

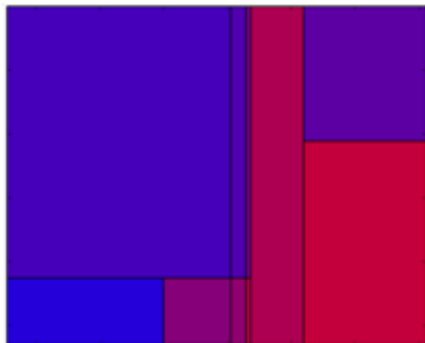




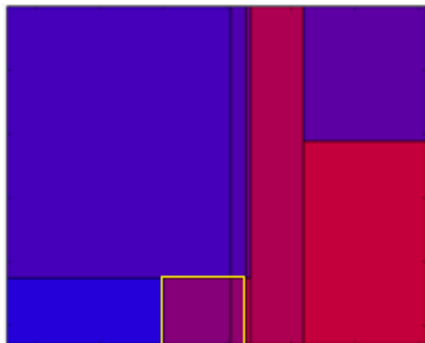
# Merging decision spaces



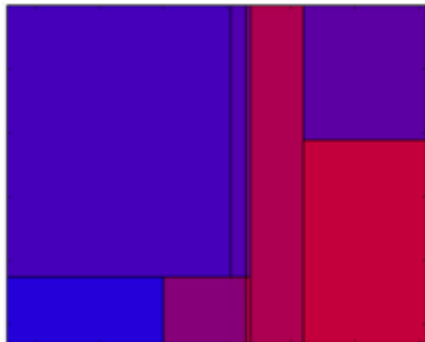
# Merging decision spaces



# Pruning decision spaces



# Pruning decision spaces

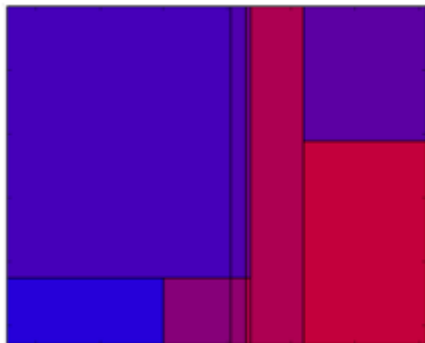


## Decision space $\rightarrow$ decision tree

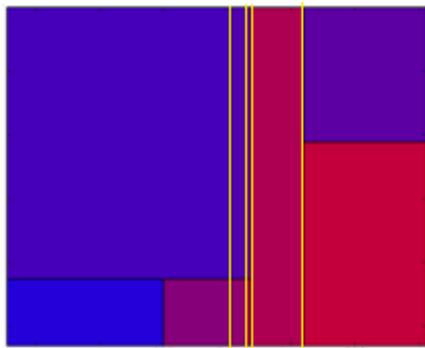
### Converting decision spaces to decision trees

One-to-one mapping from decision tree to space is lost during conversion because the order is lost. Therefore, a **heuristic** approach must be taken, identifying **hyperplane candidates** and calculating a metric to choose the 'best' plane.

## Decision space $\rightarrow$ decision tree



## Decision space $\rightarrow$ decision tree



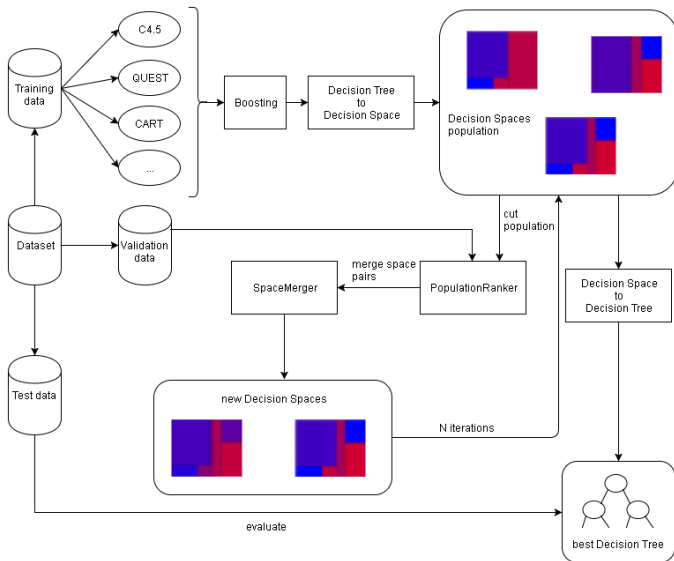
## Decision space $\rightarrow$ decision tree

### Finding 'best' candidate hyperplane

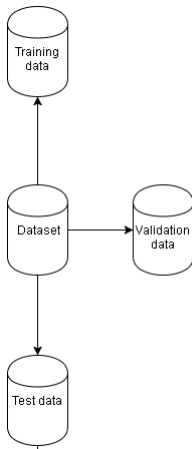
Apply metric function to each plane, these include:

- ▶ information gain and Gini
- ▶ pick plane from most correlated feature
- ▶ pick plane that divide space in two most equal subspaces
- ▶ combination

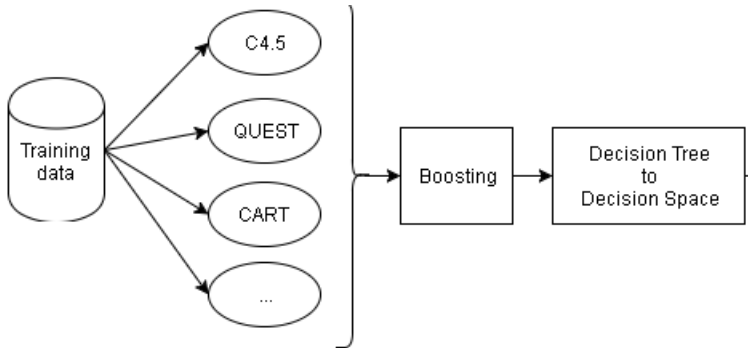




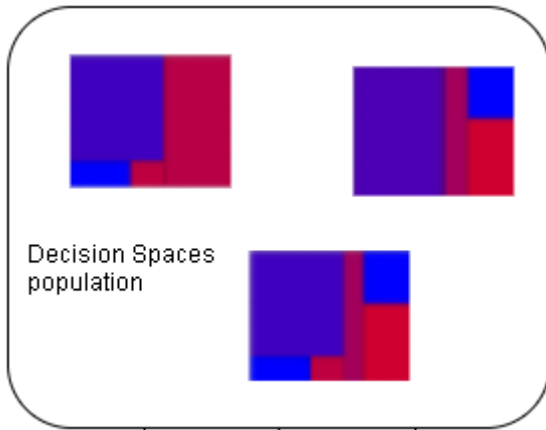
## Splitting the data



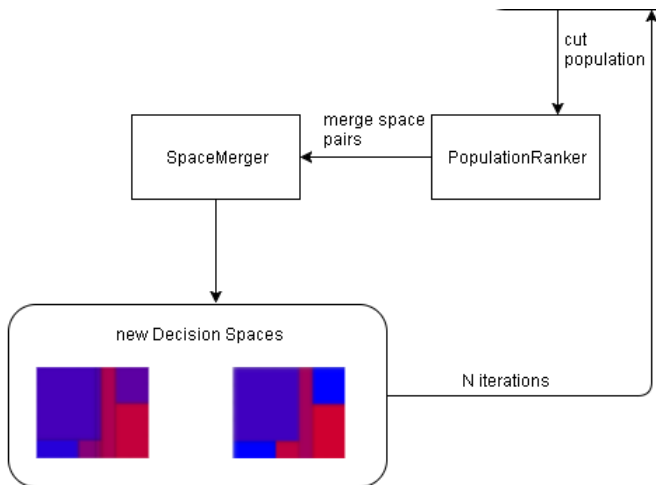
## Generate different decision trees



## Generate different decision trees



## Genetic merging

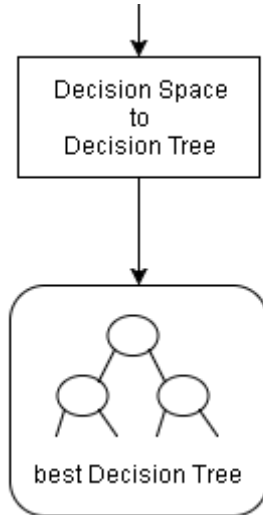


# PopulationRanker

## Fitness function

A high accuracy is the most important property of a decision tree, followed by its' size ( $\rightarrow$  comprehensibility). Genetic algorithms are well suited for **multi-objective optimization**.

## Final iteration



## Evaluating our algorithm

5 datasets from UCI

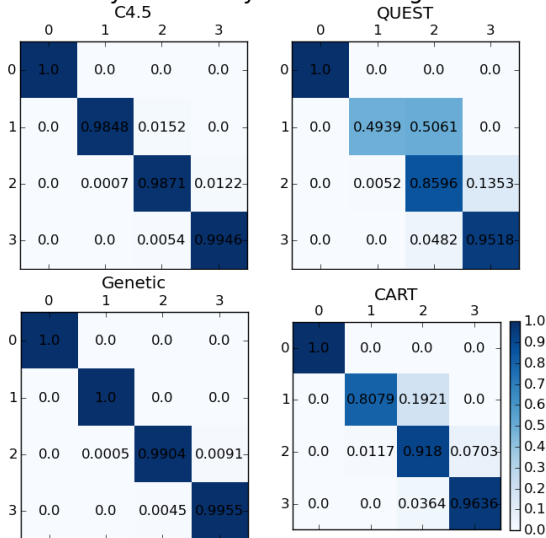
optimal parameters, feature selection when needed and k-fold CV

Name	#Samples	#Disc	#Cont	#Class	Imbalance rate
Heart	270	7	6	2	0.058
Car	1728	6	0	4	0.225
Iris	150	0	4	3	0
Shuttle	14500	0	9	7	0.18308
Nursery	12960	8	0	5	0.1498

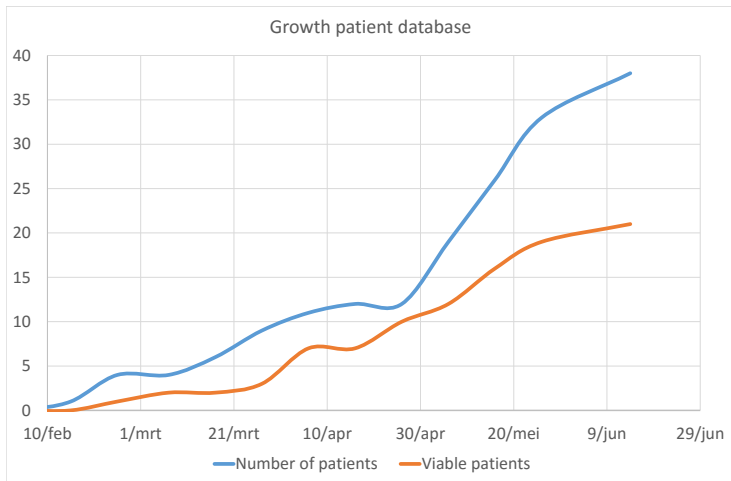


Dataset	Folds	C4.5	CART	QUEST	Genetic
Heart disease	5	<u><b>0.8067</b></u>	0.7844	0.7844	<u><b>0.8067</b></u>
	10	<u><b>0.8104</b></u>	0.7732	0.7881	0.7993
Iris	3	0.9533	0.9467	0.9467	<u><b>0.96</b></u>
	5	0.9467	0.9333	0.9467	<u><b>0.9533</b></u>
Cars	3	<u><b>0.9722</b></u>	0.9693	0.9229	0.9693
	5	0.9711	0.9682	0.9241	<u><b>0.9786</b></u>
	10	0.9756	0.9751	0.9265	<u><b>0.9803</b></u>
Shuttle	3	0.9987	0.9983	0.9964	<u><b>0.9988</b></u>
	5	0.9986	0.9981	0.9962	<u><b>0.9988</b></u>
	10	0.9990	0.9987	0.9941	<u><b>0.9992</b></u>
Nursery	3	0.9890	0.9431	0.9147	<u><b>0.9914</b></u>
	5	0.9918	0.9498	0.9251	<u><b>0.9958</b></u>
	10	0.9937	0.9568	0.9259	<u><b>0.9954</b></u>

## Accuracy on nursery dataset using 10 folds



## Headache dataset



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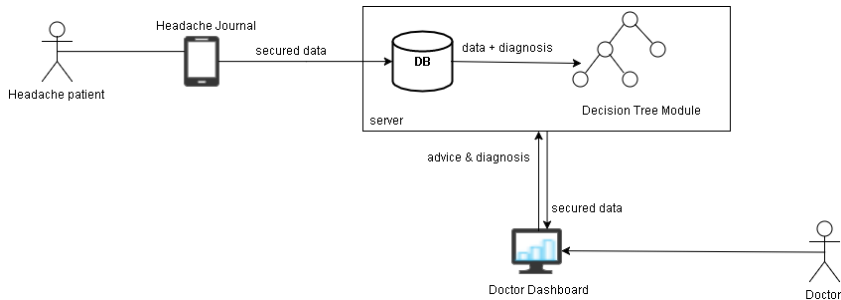
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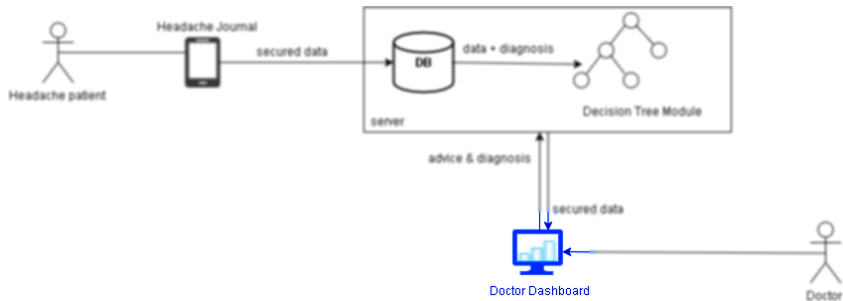
Conclusion & future work

Backend and data exposure

# Doctor dashboard



# Doctor dashboard



# Doctor Dashboard

- ▶ Web application in order for the doctors to access the data exposed by our REST API
- ▶ Preferably in the form of visualizations, which allow to process a lot of data in a small amount of time
- ▶ Developed by Maarten Vanden Berghe

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## Conclusion

The current process in the UH of Ghent can be completely digitized:

- ▶ Collect information using a mobile application
  - More efficient than paper calendars
- ▶ Present the data through a web application
  - Visualizations allow to process a lot of information quickly

This leads to an increased efficiency and reduced frequency of consults, resulting in lower health care costs.

## Conclusion

A new ensemble technique was developed and tested on very varying datasets:

- ▶ increases classification performance
- ▶ preserves excellent interpretability in contrast to current ones  
→ Can easily be visualized

When a higher number of headache patients are registered in our system, the resulting decision tree can be used to support physicians in forming a diagnosis

## Future work

- ▶ Develop native applications for iOS and Android to enhance look-&-feel
- ▶ Re-evaluate our machine learning models on a larger headache dataset
- ▶ Implement more induction algorithms and ensemble techniques to create a more diverse initial population
- ▶ Experiment with other selection techniques and fitness functions
- ▶ Optimize the heuristic approach to convert decision spaces to decision trees

# Thank you for your attention!

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