

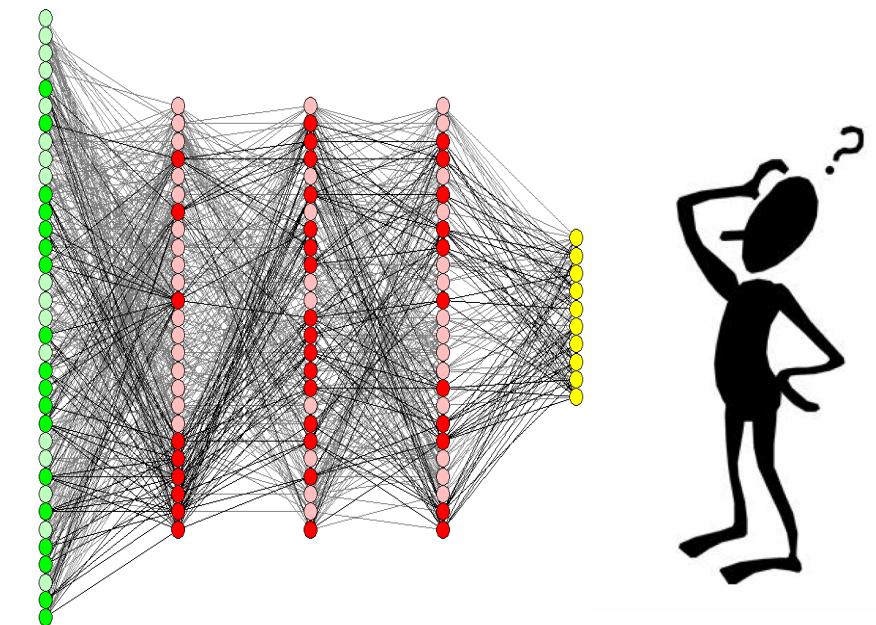
Understandable Explanations for Black-Box Models

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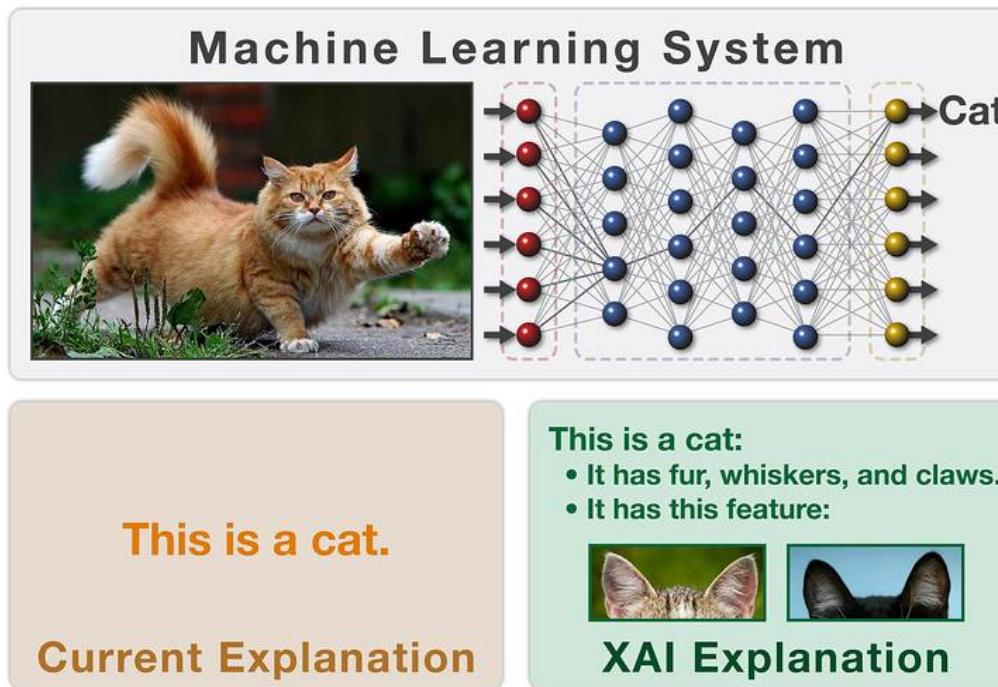
Work with Roberto Confalonieri, Fermin Moscoso, Tarek Besold

AI in Finance

- ▶ Recent progress: **AI for Finance with Deep Learning**
 - ▶ **Prediction** – prices, performance, trends
 - ▶ **NLP** – from sentiment analysis to chatbots ...
 - ▶ **Modelling customers** – credit, fraud, design, marketing
 - ▶ Possible because of **Big Data**
 - ▶ Social media, mobile devices
 - ▶ Remote sensing, image analysis ...
- ▶ Problem: Deep Learning is a **Black Box**
 - ▶ for investors, banks, insurers, regulators, developers, scientists



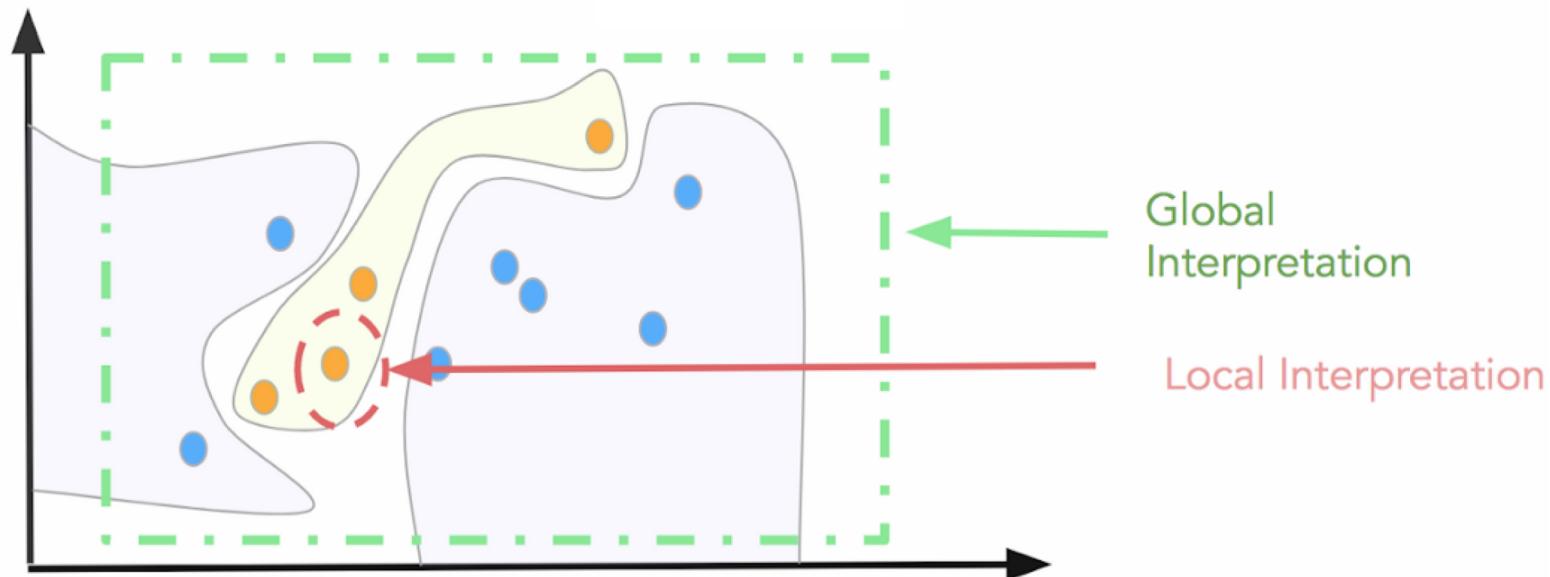
Explaining Black Box Models



- ▶ Need **transparency** and **scrutiny** for
 - ▶ performance, robustness, efficiency, ethics, fairness
- ▶ **DARPA XAI Program (2016)**
- ▶ **GDPR** Customer right for explanation (2018)

Explaining Black Box Models

- ▶ **Built-in:** combine performance and interpretability (rare)
- ▶ **Post-hoc:** Train model and explain afterwards (surrogate)
- ▶ **Global:** represent the whole model
- ▶ **Local:** focus on a single item



Understandable Explanations

- ▶ **Explanations** should be
 - ▶ accurate
 - ▶ actionable
 - ▶ **understandable**

- ▶ Focus here on **understandability**
 - ▶ Goal: **match human thinking better**
 - ▶ Approach: **Use background knowledge**
 - ▶ Needed: Identify **technical concepts** for understandability
 - ▶ Test: **Experiments with human subjects**



Explaining with Decision Trees

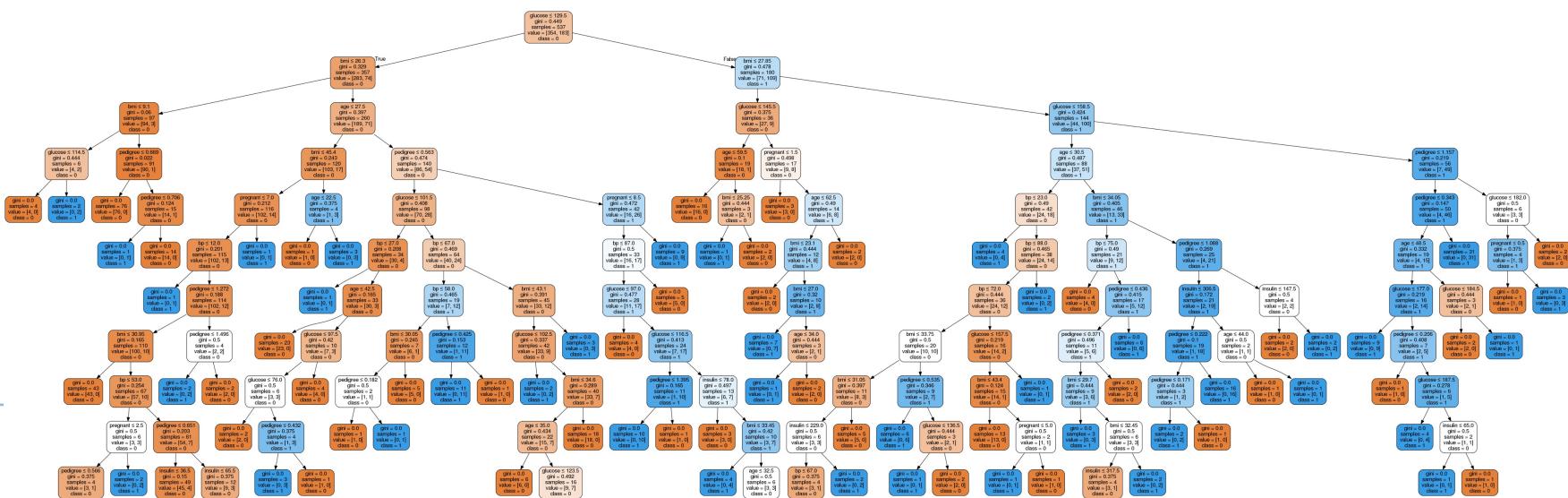
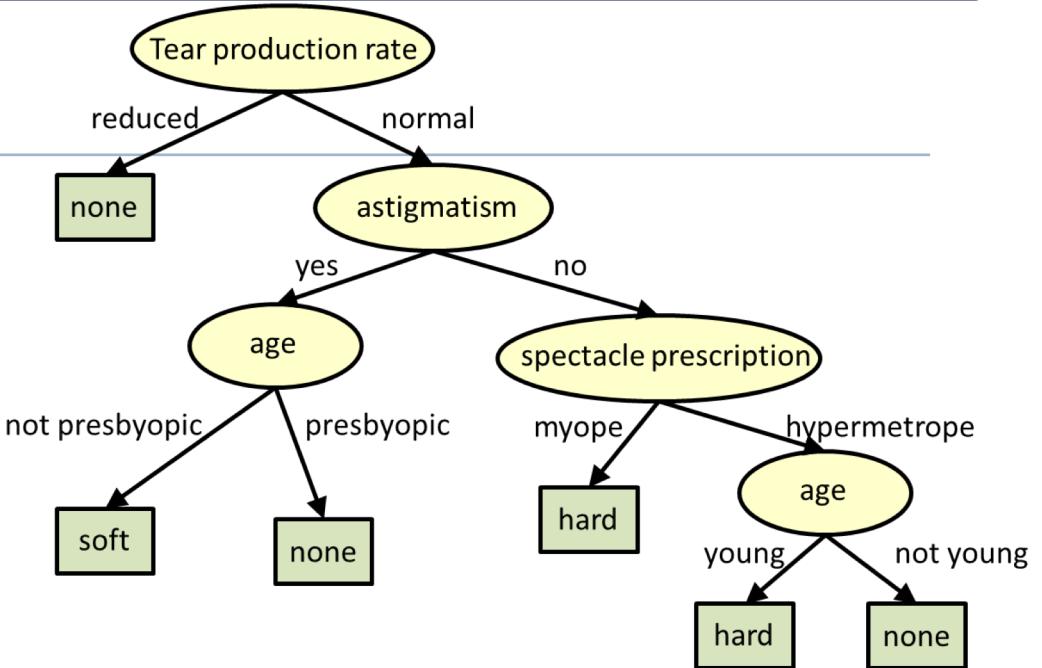
- ▶ **Global post-hoc approach:**
 - ▶ Approximate trained black box model with explainable alternative (surrogate)
- ▶ **Trepan – (Craven & Shavlik 1993)**
 - ▶ Extract Decision Trees from any black box model
 - ▶ Sample output from trained model
 - ▶ Build DT on original features
 - ▶ Typically better DT than with original data alone

Decision Trees

- ▶ DT: easy to understand
 - ▶ Elements have **meaning**
 - ▶ Readable **rules**
 - ▶ **Split feature selection by information gain**

$$\text{IG}(X_i, S) := H(S) - \sum_{j=1}^k \frac{|S_j|}{|S|} H(S_j)$$

- ▶ ... but can become unwieldy



Bringing in Background Knowledge

- ▶ **Background knowledge** modelled in **ontologies**
 - ▶ Taxonomy as **concept hierarchy** (e.g. biology, medicine, library catalogues, product catalogues)
 - ▶ **Ontology** adds **logic** (constraints, additional information, e.g. family tree, Gene Ontology, general knowledge: DBpedia, SUMO)
- ▶ **Hypothesis: general concepts are easier to process** for humans
 - ▶ Quantified as **Information Content IC**

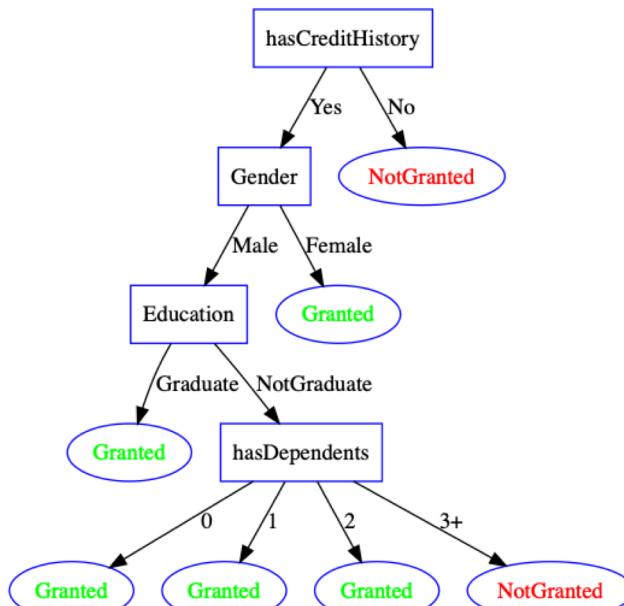
$$\text{IC}(X_i) := 1 - \frac{\log(|\text{subConcepts}(X_i)|)}{\log(|\text{sub}(\mathcal{T})|)}$$

Trepan Reloaded: Idea

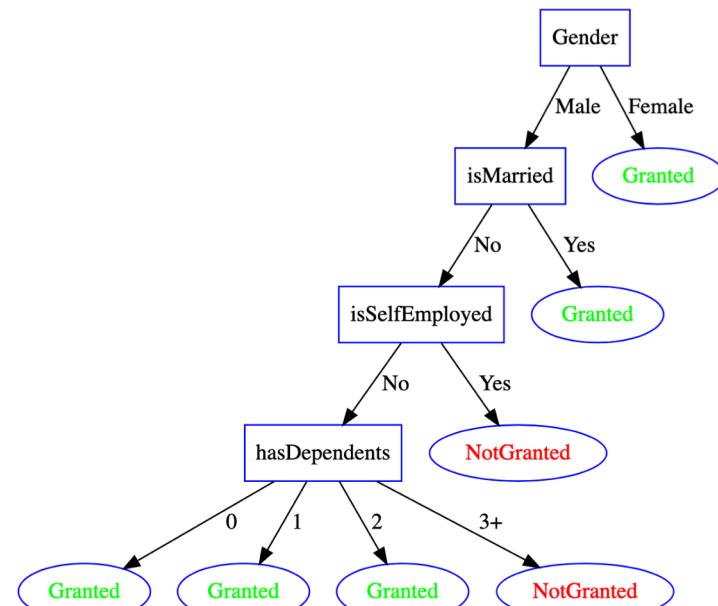
- ▶ Prefer general features for split nodes
- ▶ New reward function for selecting split features:

$$\text{IG}'(X_i, S | \text{IC}) := (1 - \text{IC}(X_i))\text{IG}(X_i, S)$$

Trepan



Trepan Reloaded



Evaluation

- ▶ **User experiment**
 - ▶ Human performance and subjective ratings
 - ▶ Online, 63 subjects, age 33 (± 12.23), 46f/17m
- ▶ **Datasets**
 - ▶ **Finance:** Kaggle Loan Dataset (selected by 34 subjects)
 - ▶ **Medical:** Cleveland Heart Disease Data (selected 29 subjects)
- ▶ **Technical factors**
 - ▶ Tree ***syntactic complexity*** for n leaves and b branches

$$\alpha \frac{n}{k} + (1 - \alpha) \frac{b}{k^2}$$

Experiments I

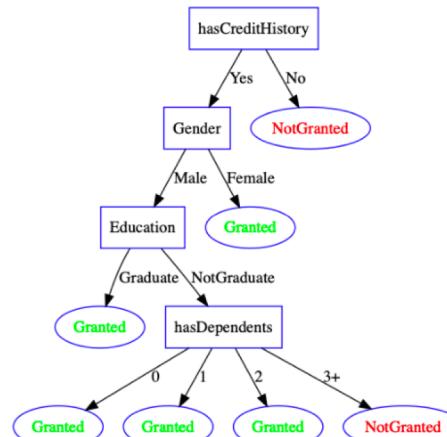
- ▶ Classification task
- ▶ Determine class of given case with DT
- ▶ 6 samples (2 each small/medium/large)
- ▶ Also rating confidence and understandability

EXAMPLE OF CLASSIFICATION TASK

In the classification task you will be asked to classify an example using a classification tree that will be shown to you.
Please have a look at this page and familiarize yourself with the task and the questions. The following pages will follow a similar pattern.

Classify the example at the top using the classification tree at the bottom.

Attribute	Value
Gender	Female
isMarried	No
hasDependents	0
Education	Graduate
isSelfEmployed	No
ApplicantIncome	3510
CoApplicantIncome	0
hasLoanAmount	76
hasLoanAmountTerm	360
hasCreditHistory	No
PropertyArea	Urban



1. The example is classified as / belongs to the class:

- Granted
- NotGranted

Experiments 2

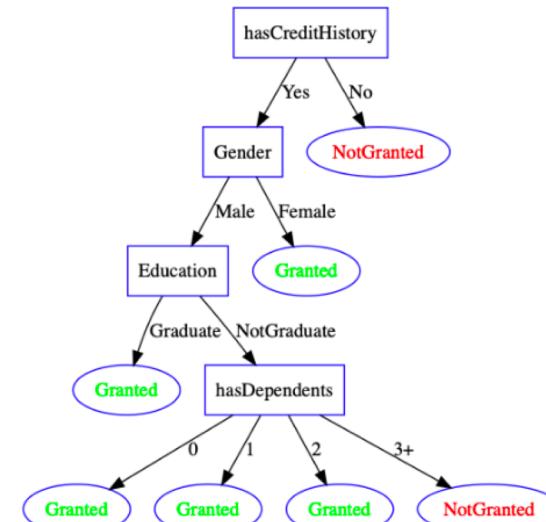
- ▶ **Inspection task**
- ▶ **Determine if a given statement is true in a DT**
- ▶ **6 samples (2 each small/medium/large)**
- ▶ **Also rating confidence and understandability**

EXAMPLE OF INSPECTION TASK

In the inspection task you will be asked to tell whether a sentence describing (part of) a classification tree is true or false.
Please have a look at this page and familiarize yourself with the task and the questions. The following pages will fo

Is the **following statement true or false** with respect to the **classification tree** shown below?

You are a female; your level of education can affect the decision outcome.

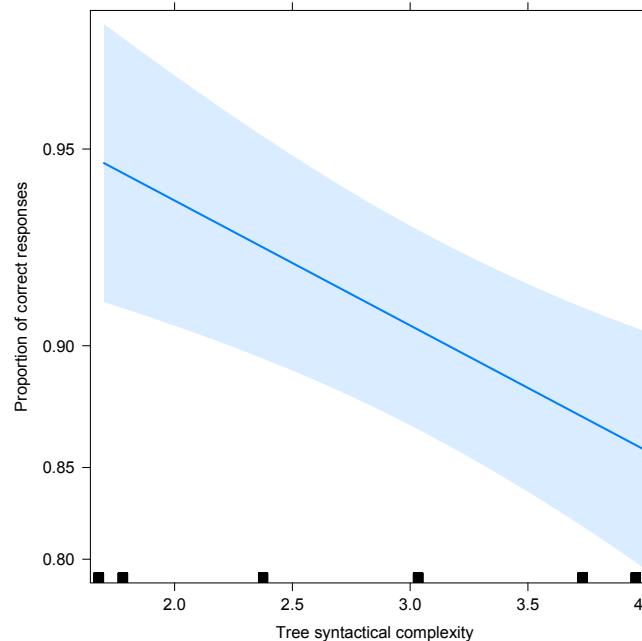
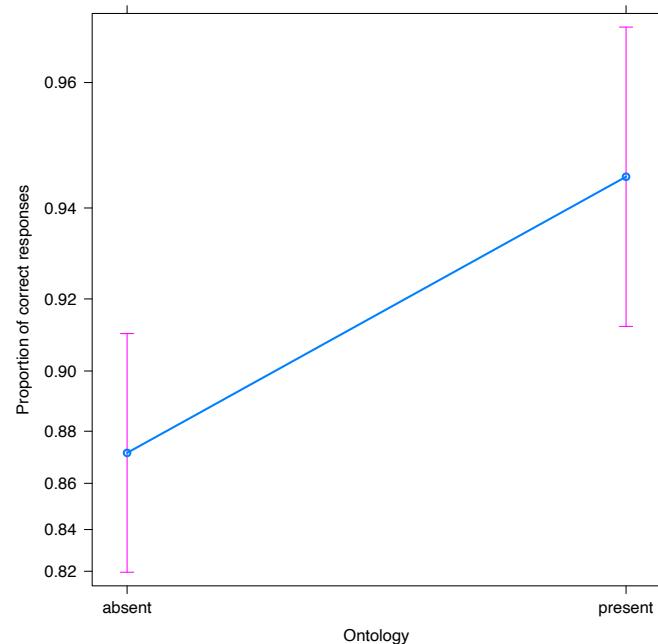


1. The above statement is:

- True
- False

Results Experiments I & 2

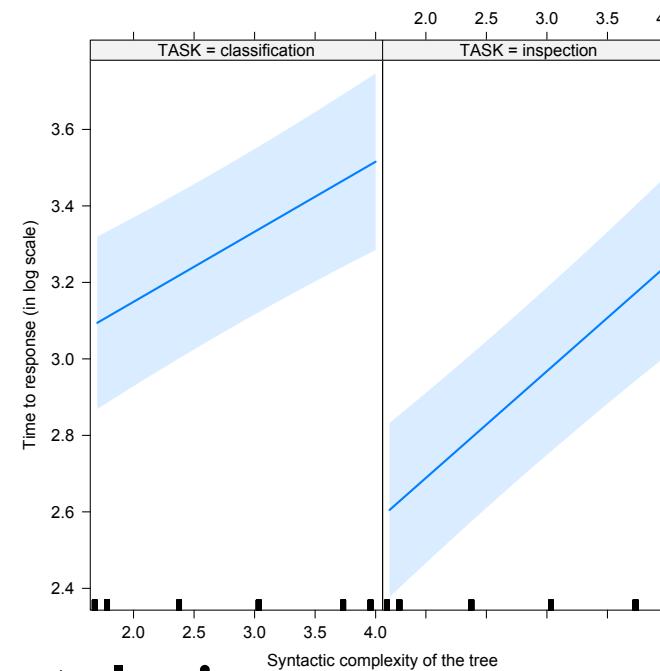
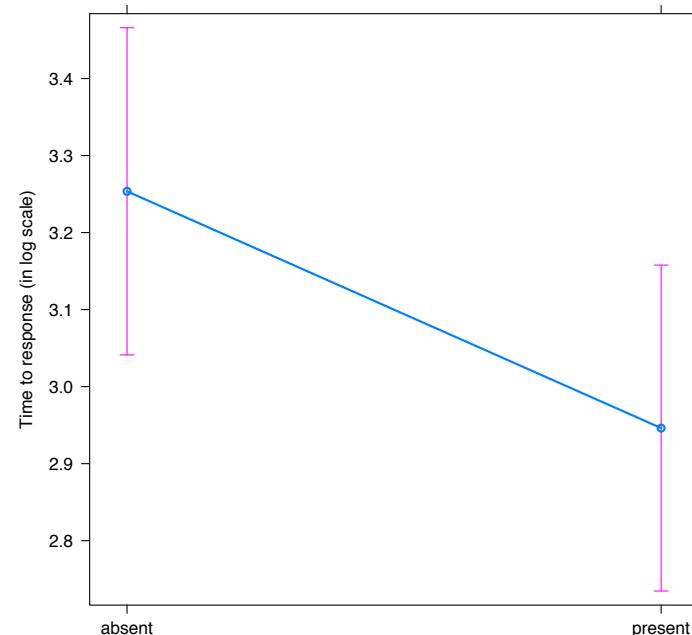
- ▶ **Correctness of responses:** mixed-effect logistic regression (task, ontology, syntactical complexity)



- ▶ Responses with ontology are more often correct
- ▶ Higher syntactic complexity makes task harder
- ▶ Subjective understandability very similar
- ▶ All effects significant at $p < 0.01$

Results Experiments I & 2

- ▶ Response **time** (correct responses): mixed-effect linear regression (task, ontology, syntactical complexity)



- ▶ Correct responses ^{ontology} quicker with **ontologies**
- ▶ Higher **syntactic complexity** makes task **harder**
- ▶ **Confidence** results similarly significant
- ▶ All effects **significant** at $p < 0.01$

Experiments 3

- ▶ **Comparison task**
(subjective)
 - ▶ Rating **which DT** is more
understandable
 - ▶ 3 samples
 - ▶ **Result:** DT with
ontology subjectively
more understandable
($p < 0.01$)

Which tree is **more understandable**?



- 1. Select the statement that best fits your opinion:**

- The tree at the top is much more understandable
 - The tree at the top is more understandable
 - The trees at the top and at the bottom are equally understandable
 - The tree at the bottom is more understandable
 - The tree at the bottom is much more understandable

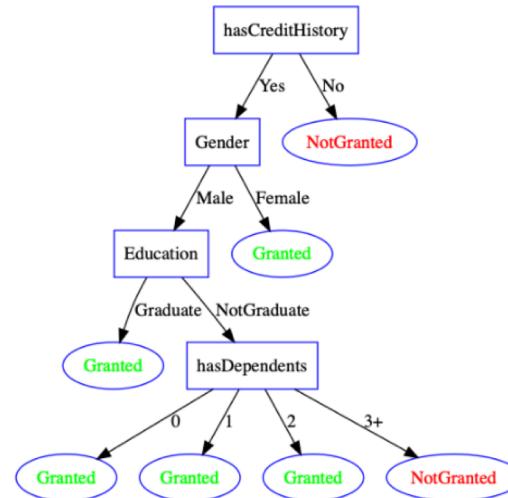
Experiments 4

- ▶ Empowerment (actionability)
- ▶ Determine how you could **change** the **outcome** for a given case with a given DT

EXAMPLE OF EMPOWERMENT TASK

In the empowerment task you will be asked to specify what event or what action you could take according to the decision tree to change a decision outcome.
Please have a look at this page and familiarize yourself with the task and the questions. The following pages will follow a similar pattern.

Specify what event could change the decision outcome.
In providing the answer, notice that you **can also change the premises provided**.



Please have a look at the question and answer, you do not have to provide any answer here.

1. You are a male and you have 3 children: is there an action (only 1) that you could take according to the decision tree to become eligible for the loan? [You can also change these premises].

Yes

No

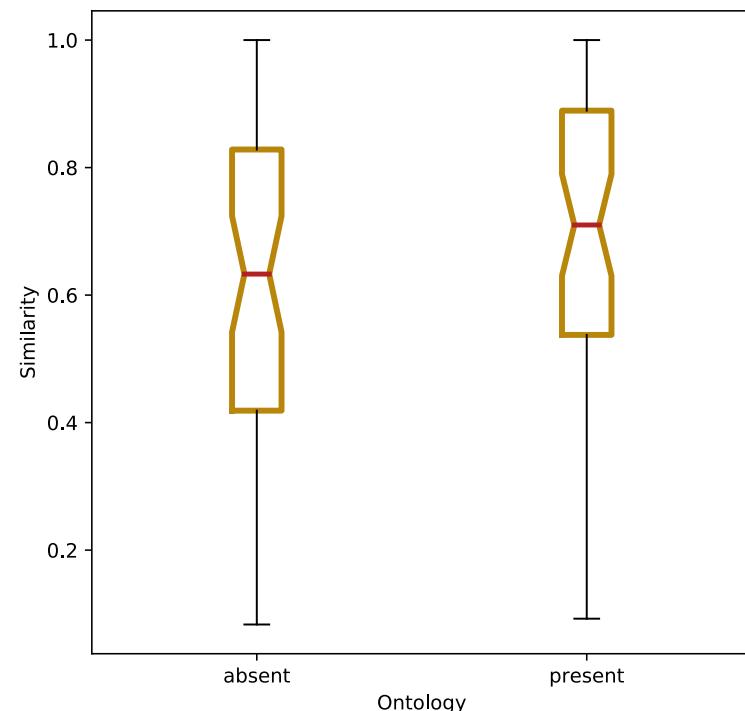
Not applicable

2. If yes, what is such an action/event?

The action/event is: become a graduate

Results Experiments 4

- ▶ Evaluation
 - ▶ Semantic similarity of free text answers
 - ▶ SpaCy with pre-trained word embeddings
 - ▶ Analysed for similarity with pre-defined answers
- ▶ Ontology significantly increases correctness and reduces response time



Conclusions

- ▶ **Explanation generation with Trepan Reloaded**
 - ▶ Integrate semantic background knowledge
 - ▶ Hypothesis: general concepts are more understandable
- ▶ **Experimental results**
 - ▶ Human performance and subjective understandability improved significantly in all tasks
 - ▶ Hypothesis robustly confirmed
- ▶ **Ontologies makes Decision Trees more effective for human use**
- ▶ **Future work**
 - ▶ Apply with more ontologies and different use cases
 - ▶ Automate ontology selection and mapping
 - ▶ Fine tune understandability for different domains and tree structures

Thank you!