Presentation

A Study on Mitigating Hard Boundaries of Decision-Tree-based Uncertainty Estimates for AI Models

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Model-agnostic Uncertainty Wrappers provide dependable uncertainty estimates Motivation

- The outcomes of Data-Driven Models (DDMs) cannot be assumed to be always correct
 - The outcomes of DDMs are subject to uncertainty
- Different approaches exist to provide 'inside-model' or 'outside-model' uncertainty estimates
- An example of the latter is the concept of model-agnostic Uncertainty Wrappers
 - Uncertainty Wrappers enrich DDMs with a dependable uncertainty estimate





Decision Trees can be used to obtain interpretable uncertainty estimates Motivation

- Uncertainty Wrappers cluster inputs with similar uncertainties based on influence factors using Decision Trees
- Benefit of Decision Trees: uncertainty estimates are easy to interpret



• Example: x_1 : semantic_visibility = 0.3, ..., *distance* = 11.450 => **uncertainty estimate** = 16%



Decision Trees may lead to undesired hard decision boundaries Problem Statement

- Problem: The discrete approach of Decision Trees realizes hard decision boundaries
- For continuous features, these boundaries may be rather unintuitive





Goals to be met by the softening approaches

Solution search

- Approaches need to satisfy the following goals:
 - G1: Implement soft decision boundaries
 - G2: Interpretable uncertainty estimates
 - G3: Reasonable runtime complexity
 - G4: Reasonable uncertainty estimation performance



- Identified approaches:
 - Random Forests:

Breiman, L. 2001. Random forests. Machine learning, 45(1).

Fuzzy Decision Trees / Random Forests:

Marcelloni, F.; Matteis, A. D.; and Segatori, A. 2015. A new approach to fuzzy random forest generation. In Proceedings of Int. Conf. on Fuzzy Systems, 1-8. IEEE.

Soft Decision Trees / Bagged Soft Decision Trees:

Alpaydin, E.; Irsoy, O.; and Yildiz, O. T. 2016. Bagging soft decision trees. In Machine Learning for Health Informatics, 25-36. Springer, Cham.

Softening is achieved by using membership functions

Solution approaches

- *Main concept:* A data point has a membership degree to the leaves
- Uncertainty estimate = Weighted sum of leaf uncertainties and membership values
- Key difference between the approaches is the type of membership function used
- Simplified example of a Fuzzy Decision Tree to determine an uncertainty estimate for an input **X**:





Softening approaches evaluated on pedestrian detection use case and metrics Evaluation

- Use case: Pedestrian detection on roads
- Study execution build datasets:



Study execution – evaluate softening approaches:





Results: Softening achieved, but decreased estimation performance Evaluation

Summary of the extent to which the approaches achieve our goals:

Approach	G1: Softening	G2: Interpretability	G3: Runtime Complexity	G4: Estimation Performance
Decision Trees (baseline)		++	++	++
Random Forests	О	+	+	+
Fuzzy Decision Trees	++	+	0	Ο
Fuzzy Random Forests	++	О	-	Ο
Soft Decision Trees	++	+	0	-
Bagged Soft Decision Trees	++	О	-	-



Softening is a trade-off decision dependent on the use case

Conclusion and future work

- A general recommendation for the use of a particular approach cannot be provided
- Rather, the selection of an approach has proved to be a trade-off decision
- Based on the study results, we see **two main directions** for further work:
 - Develop specific recommendations for choosing an approach for a concrete setting
 - Modify approaches to address the observed uncertainty estimation performance limitations

