

Explainable and Reliable AI

Research Project Presentation 2

Group 11

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Problem Description

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Scenario:

Person **A** and Person **B** feel slight discomfort in the chest area, and they request **hospitalization**. (The hospital has only **1 slot** available.)

- Which patient should the hospital accept and which one to reject?
- How can the doctor provide an explainable and reliable decision?

Decision Tree

ML model that uses data features as decision rules in a tree like structure.

- Internal Nodes: Store features as decision rules
- **Branches:** Path based on feature value
- Leaf Nodes: Store prediction outcome

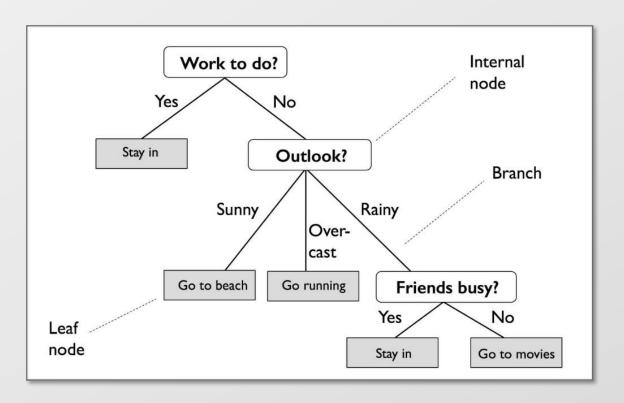


FIGURE: Decision Tree structure example

Inductive Conformal Predictor (I.C.P.)

- Returns prediction regions intervals for regression problems and sets of labels for classification problems.
- Training Calibration Test set: The percentage split of each set can differ.
- Nonconformity function: Measures how unusual an example looks relative to previous examples

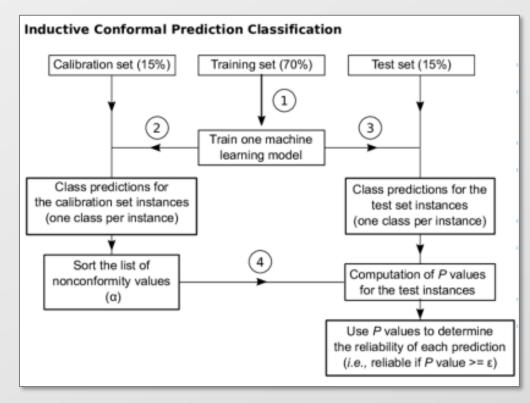


FIGURE: Inductive Conformal Predictor structure example for classification

Validity of I.C.P.

To test the validity of a conformal set predictor we use the error rate e.

The **error rate e** for a **significance level s** is defined as the proportion of test instances whose predicted prediction-sets do not contain the correct class.

To test the I.C.P. validity, we need to show that for any significance level $\epsilon \in [0, 1]$ we have $e \leq \epsilon$.

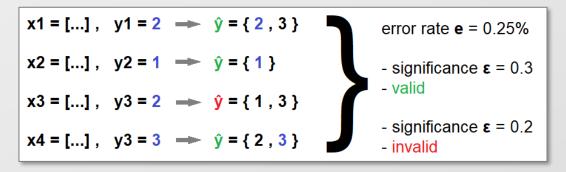


FIGURE: Validity example

Information Efficiency of I.C.P.

To test the **informational efficiency** of a conformal set predictor, for a given **significance level s** we employ **three main metrics**:

- The rate **r**^e of **empty** prediction sets
- The rate **r**^s of **single** prediction sets
- The rate $\mathbf{r}^{\mathbf{m}}$ of **multiple** prediction sets.

We need both validity and information efficiency.

Prediction sets as small as possible but NOT empty.

Our Approach

Combining Decision Tree with I.C.P.

• **Hybrid** prediction model

Normal decision tree structure

• I.C.P. at leaf nodes

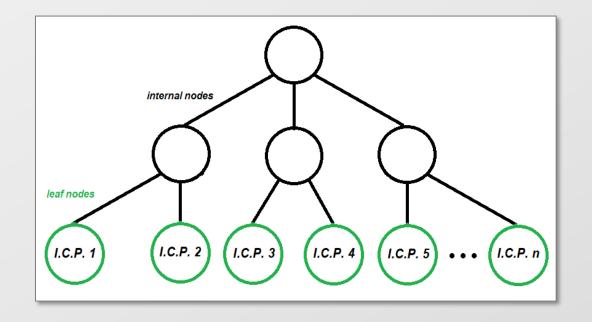


FIGURE: Decision Tree with I.C.P.s at leaf nodes

Training Process

1) **Grow tree** with training set

2) Send **training examples** through the tree

3) Send calibration examples through the tree

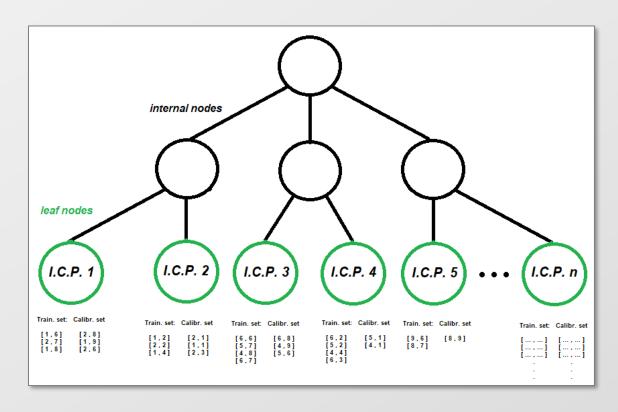


FIGURE: Training and calibration instances at leaf nodes

Classification Process

1) Send **test example** through the tree

2) Once the test instance reaches a leaf node, perform conformal prediction among those training and calibration items in the same leaf as the test instance.

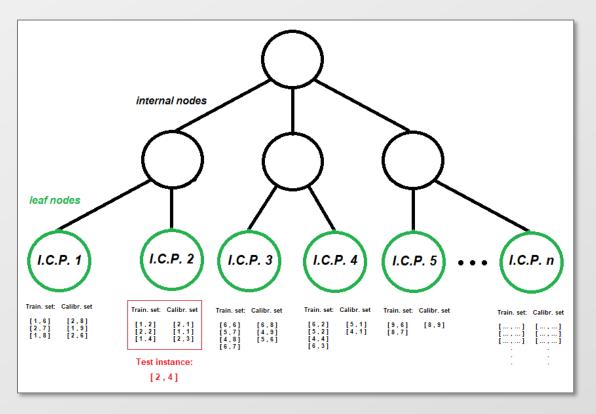


FIGURE: Leaf-specific inductive conformal predictor

First Results

Iris Dataset

First plot:

Error rate e / Significance level s

Second plot:

Mean prediction sets **size** / Significance level **ɛ**

Third plot:

Rate **r**^s of **single** prediction sets / Significance level **ɛ**

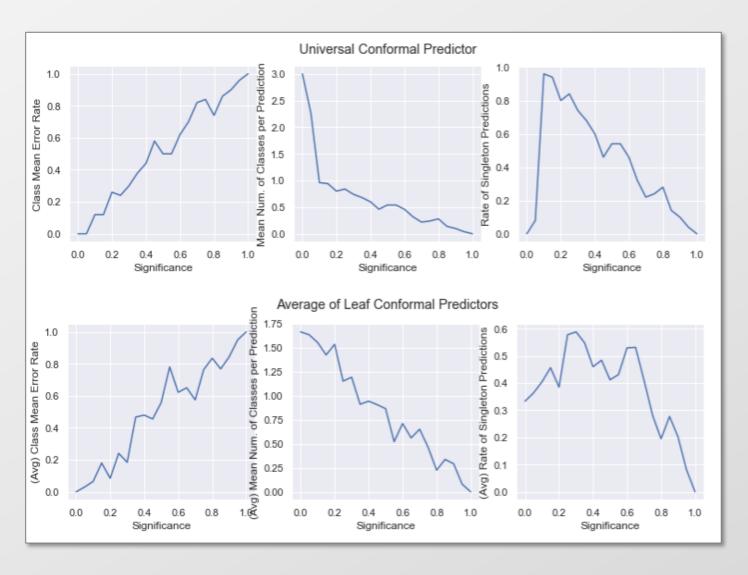


FIGURE: Plots for validity and information efficiency

Challenge

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What if there is **no calibration** or **training** set on a **leaf node**?

We can not perform I.C.P. on a test instance that arrives at such node.

Two possible solutions:

- 1) Do not perform conformal prediction at all.
- 2) Prune paths that lead to nodes with no calibration / training set.

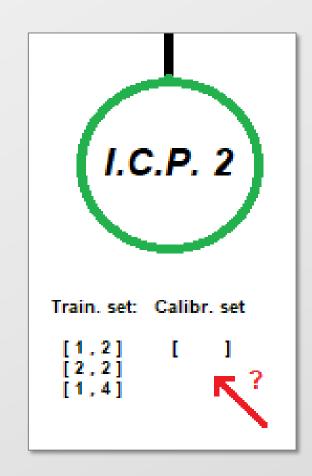
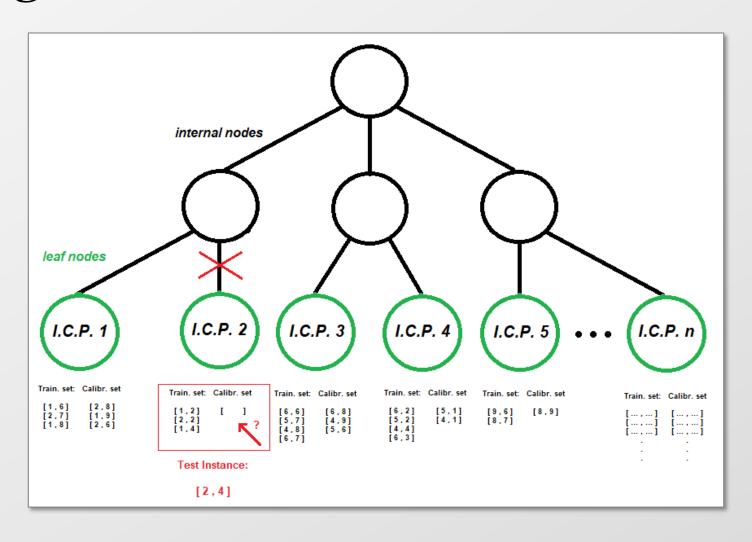
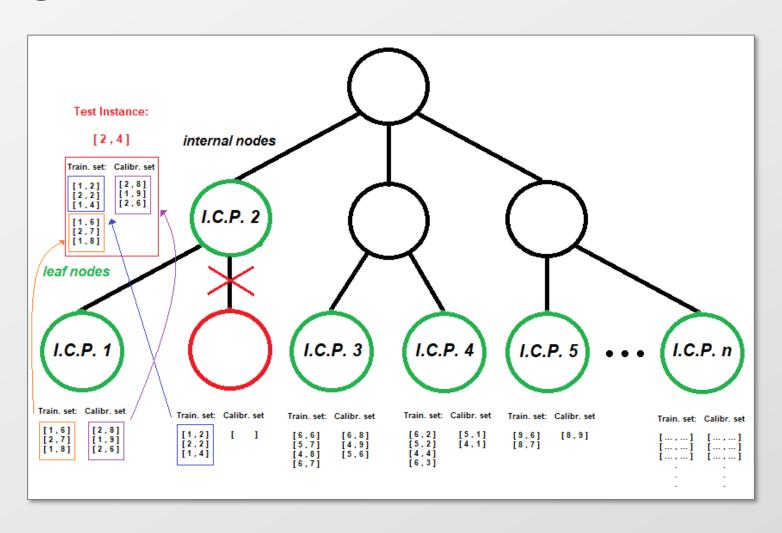


FIGURE: Leaf node with no calibration set

Pruning Method



Pruning Method



Research Questions

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• Question 1

Is **Validity** preserved?

• Question 2

Comparison of global I.C.P. performance vs average of leaf I.C.Ps performance.

• Question 3

Comparison of global I.C.P. performance vs individual leaf I.C.Ps performance.

Thank you for your attention!

Any questions?

