# Optimal Decision Trees for Interpretable Clustering with Constraints

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Constrained Clustering

Semi-supervised

**Decision Trees** 

Interpretable Classifiers

Decision Tree Clustering

No constraint support

No optimality guarantee

#### Overview



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# Problem Definition

- **Bi-criteria** objective:
  - Maximize minimum split (MS) between clusters
  - Minimize maximum diameter (MD) within clusters



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- **Bi-criteria** objective:
  - Maximize minimum split (MS) between clusters
  - **Minimize** maximum diameter (MD) within clusters
- Pairwise Constraints:
  - Must-links: pairs that should be in the same cluster
  - **Cannot-links:** pairs that should be in different clusters



# Problem Definition

• **Decision tree** clustering:







# Encoding

Legend

• All distances of pairs sorted into **distance classes** 





# Encoding

Legend

• All distances of pairs sorted into **distance classes** 





## Smart Pairs

• Linear number of clauses enough to enforce quadratic number of must-links





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	Constraints	Conditional		Detect <u>redundant</u> edges
Force to be in the same cluster	Must-link	MS obj.	Smart Pairs	
Force to be in different clusters	Cannot-link	MD obj.		Detect <b>infeasible</b> edges

# Better Score + Better Interpretability

• High quality solutions in a short time



- Improve the solution individually
- Complement each other

• Trade-off between quality and feasibility



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# Thank you for your time!

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