# Ensemble Trees

## Leveraging Ensemble Power inside Decision Trees

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Katholieke Universiteit Leuven

# The Setting

- Vectorial data $\langle v_1, v_2, v_3, \ldots, v_d \rangle$
- Binary class {pos,neg}

# Motivation

Decision trees easy to induce

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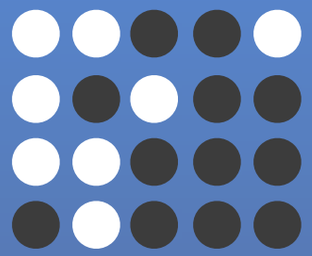

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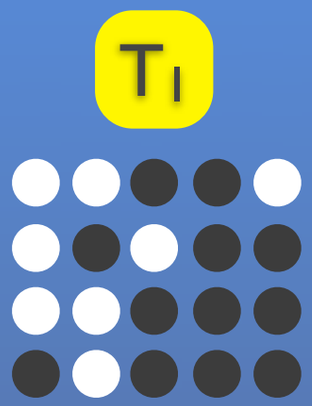

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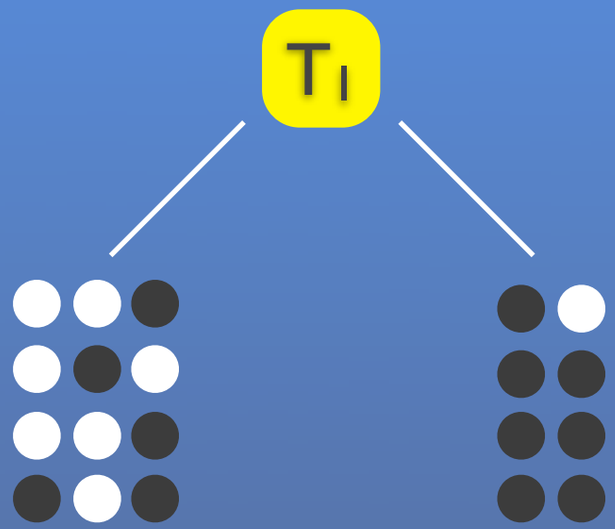

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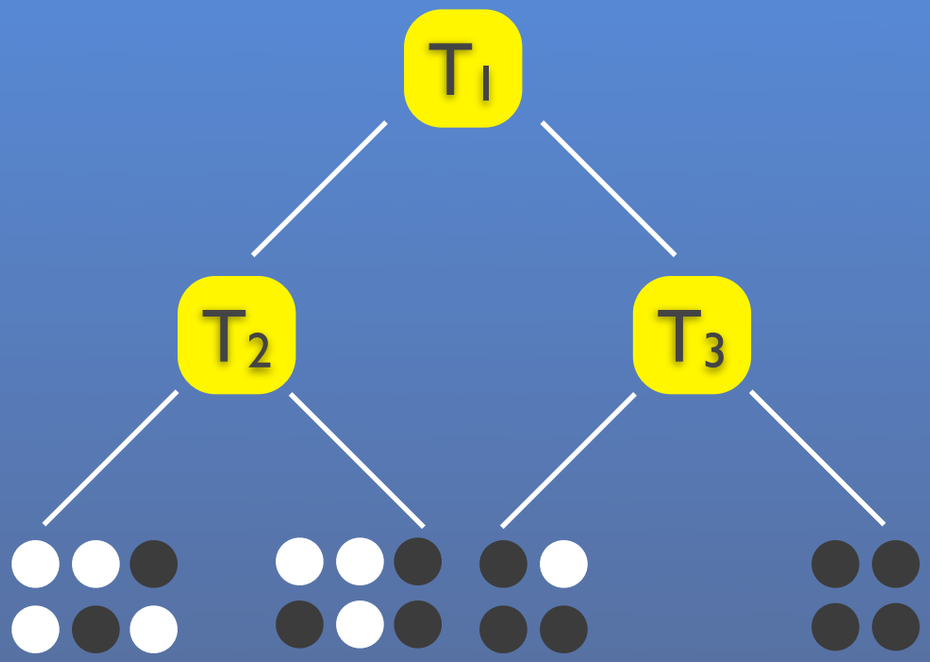

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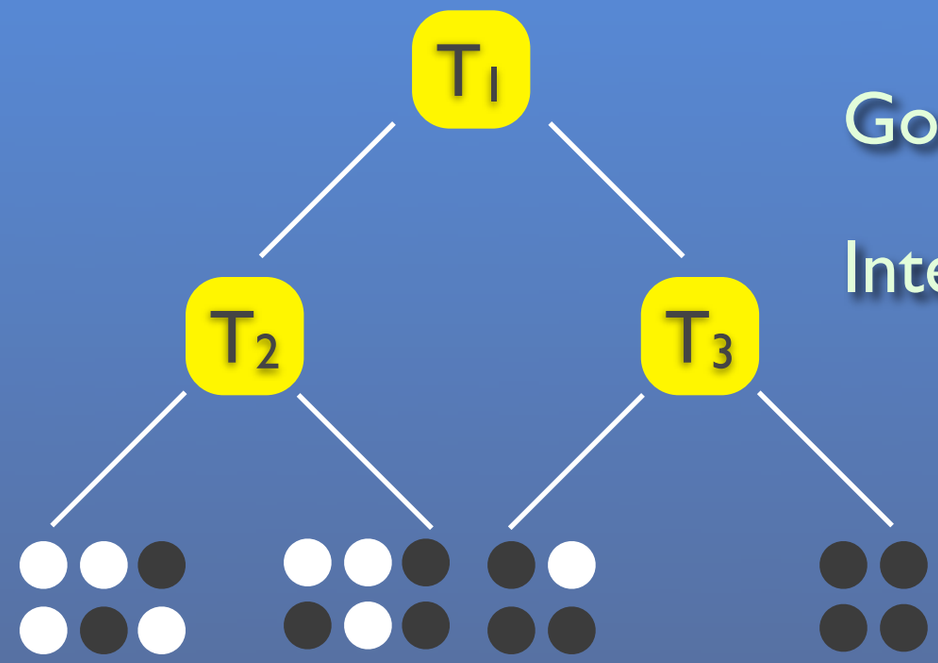


Good accuracy ✔

Interpretable ✔

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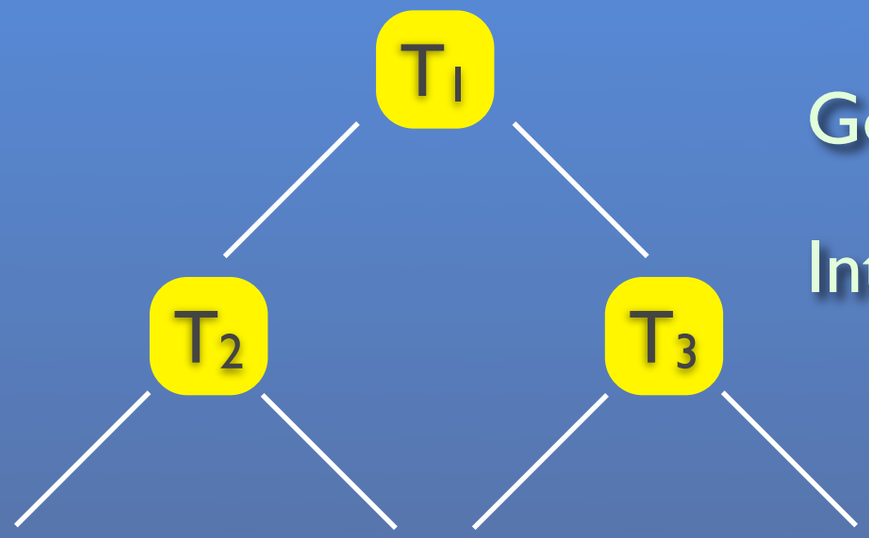


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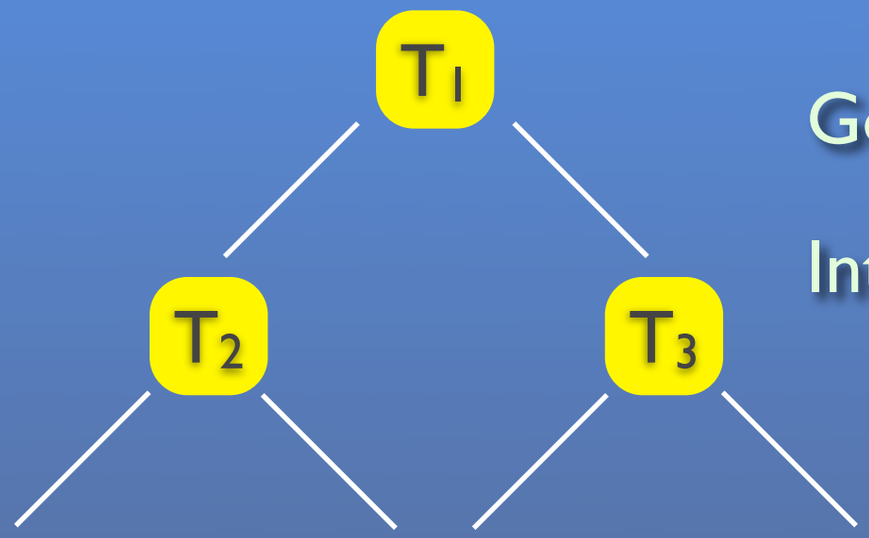


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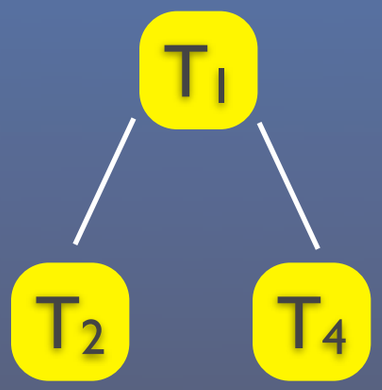

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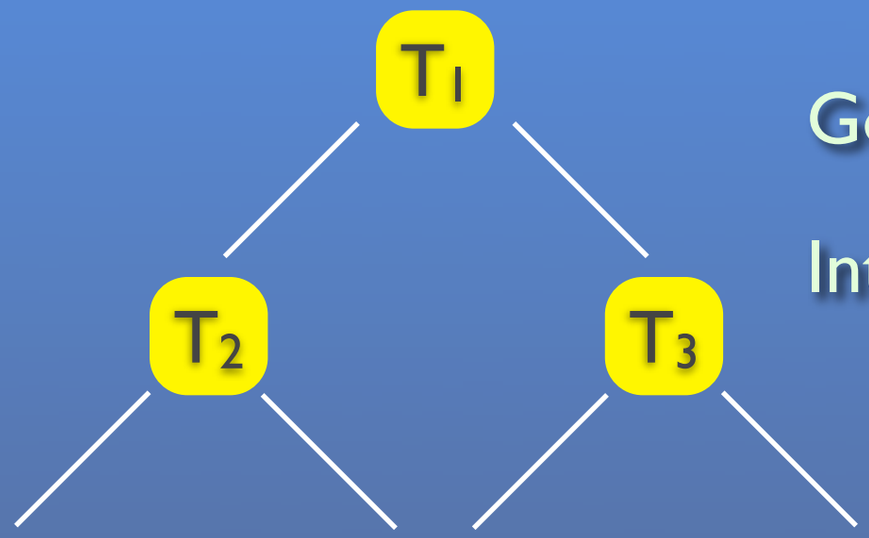


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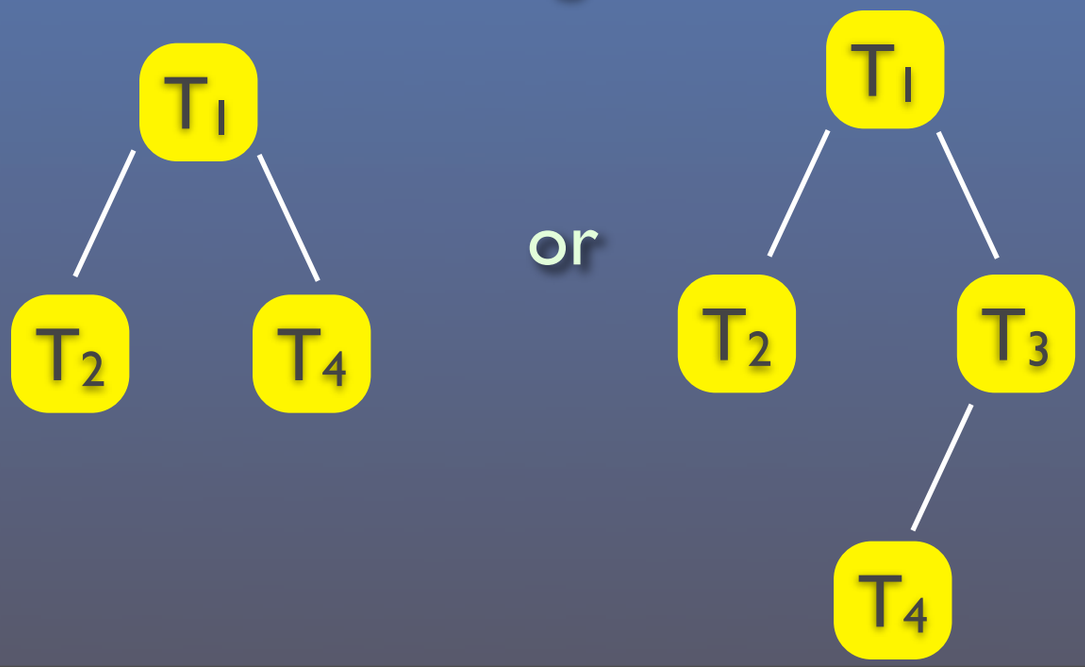

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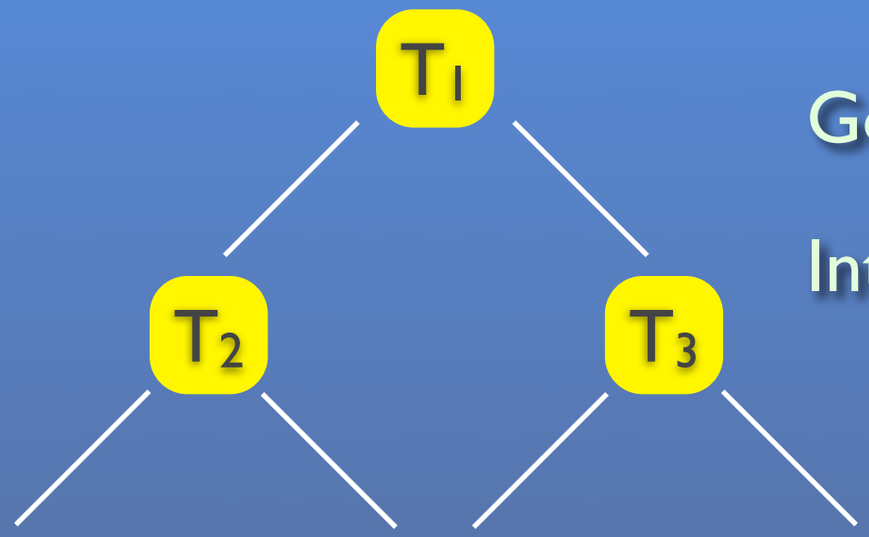


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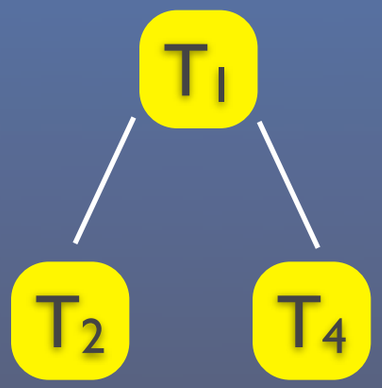


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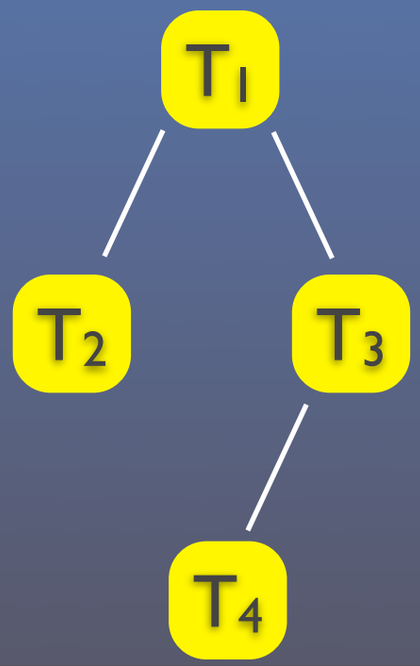


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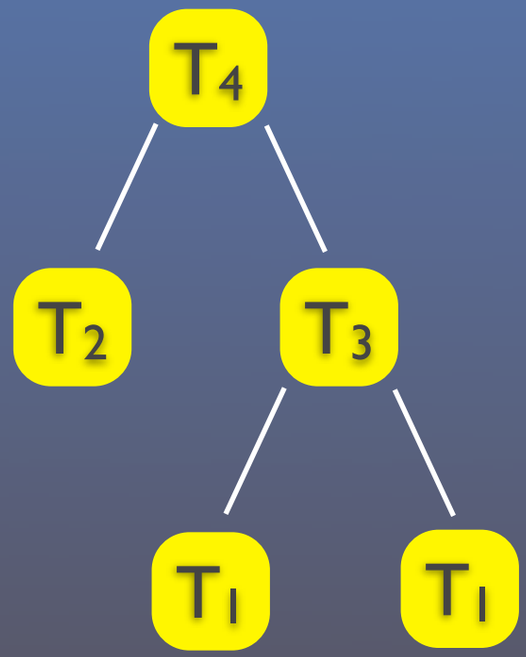

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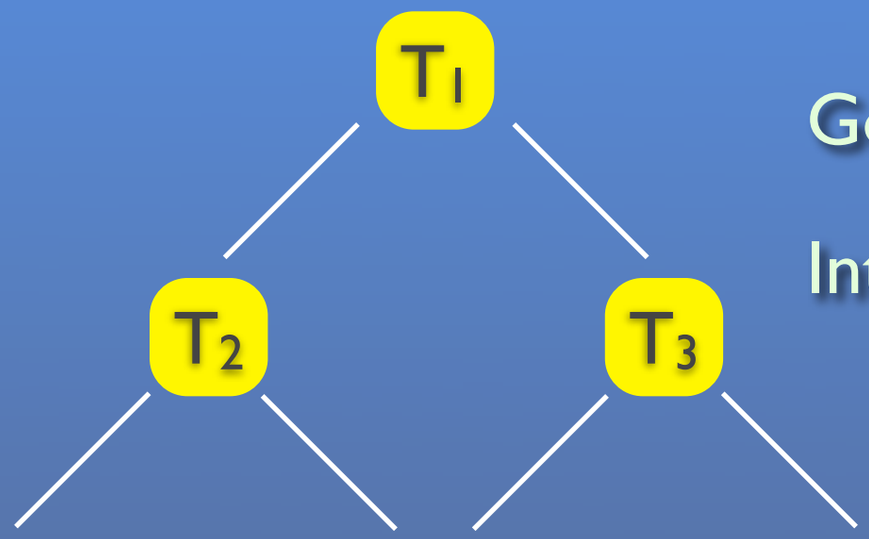


Good accuracy ✓

Interpretable ✓

Varying accuracies ✗
Faulty interpretations ✗

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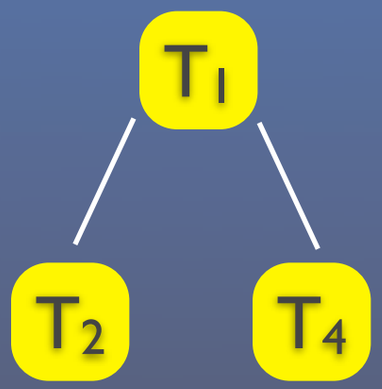


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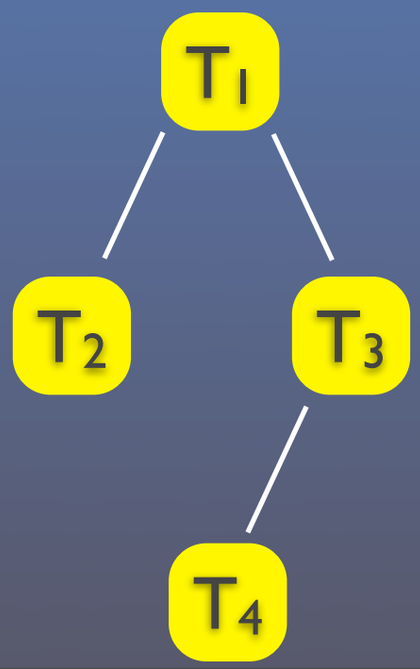


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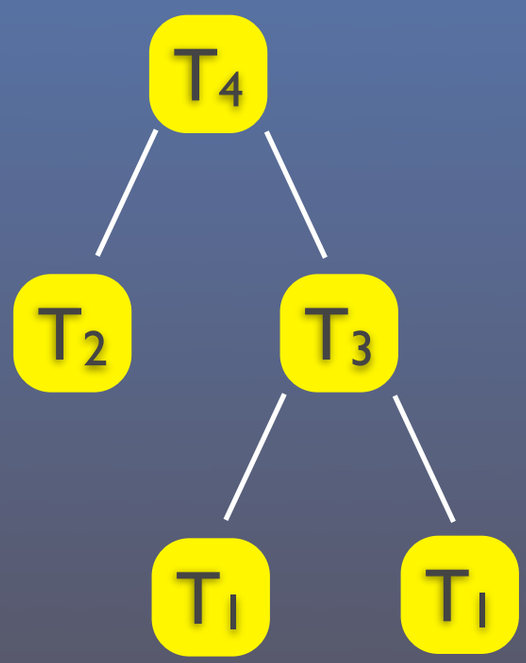

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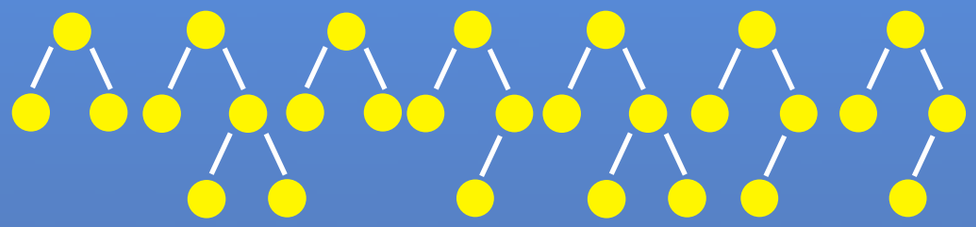

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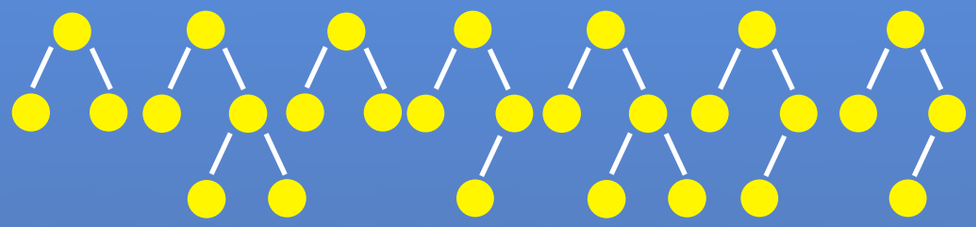

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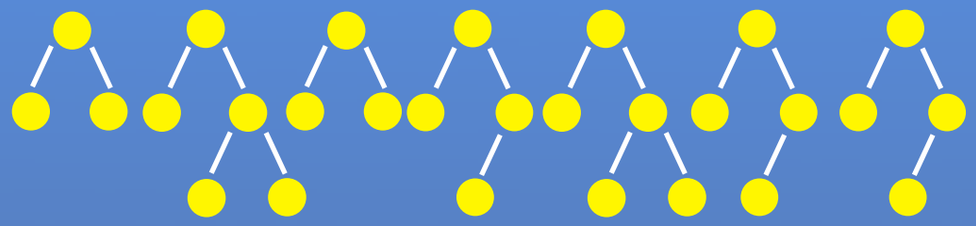

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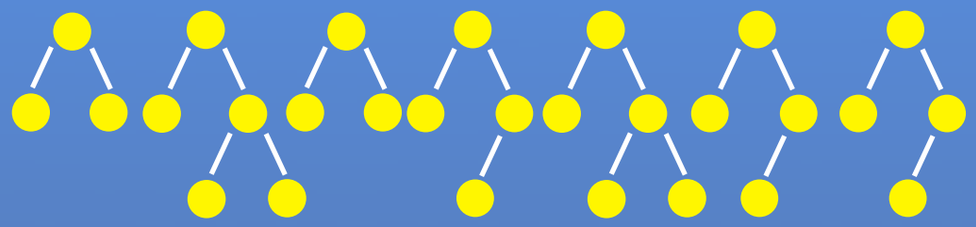

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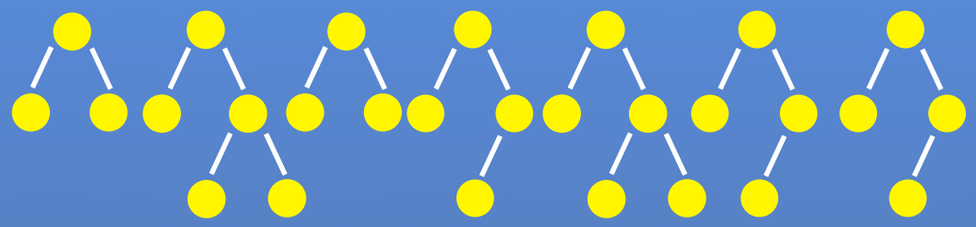

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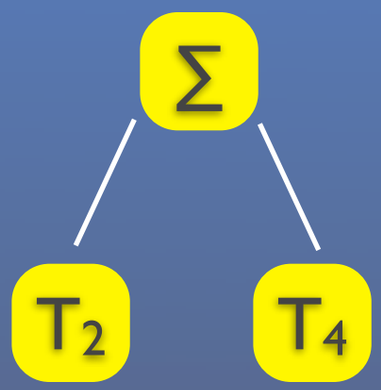


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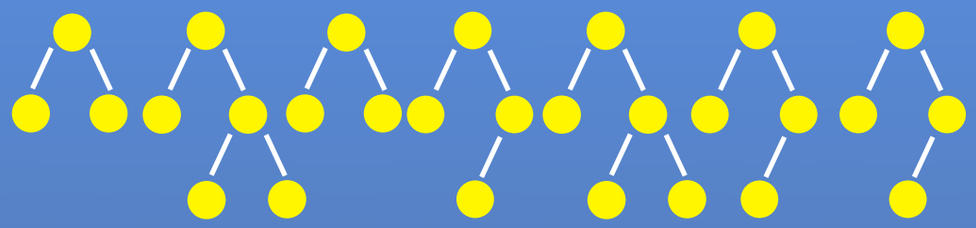

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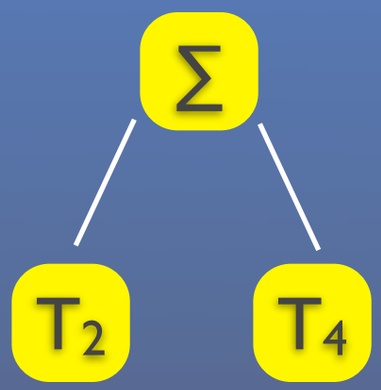


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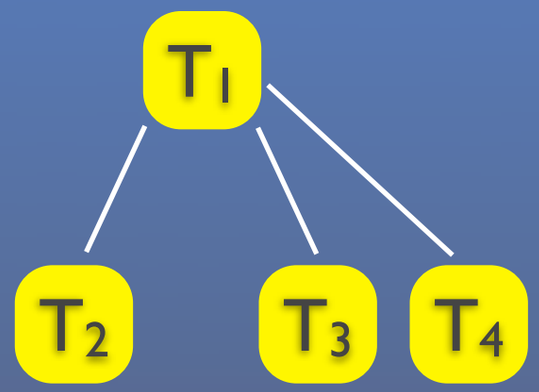


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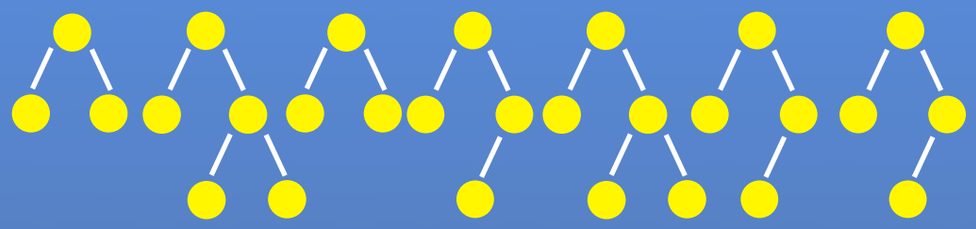

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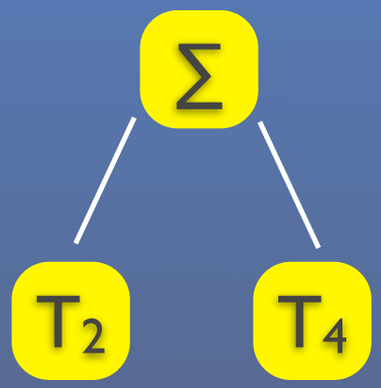


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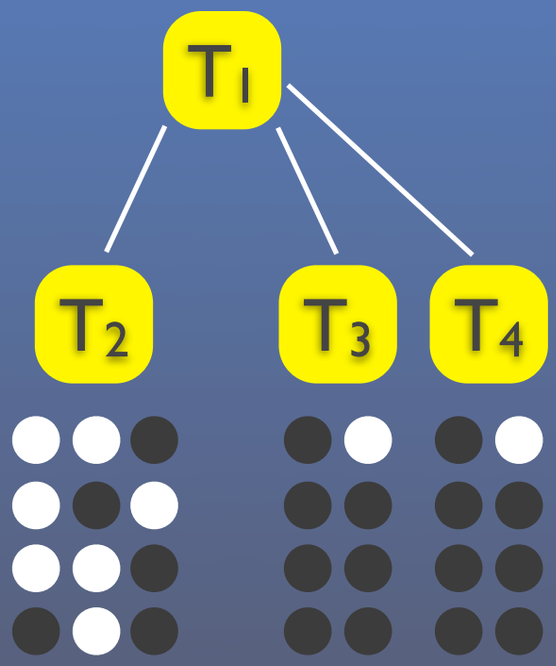


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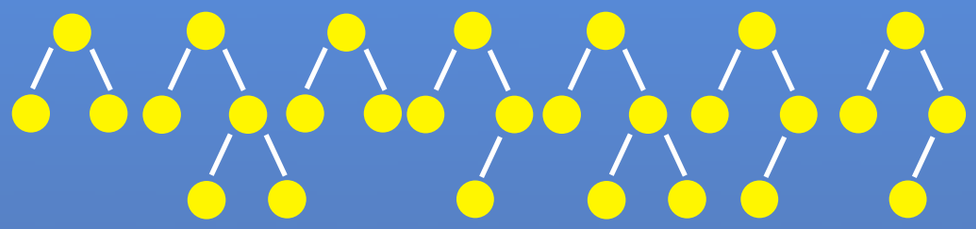

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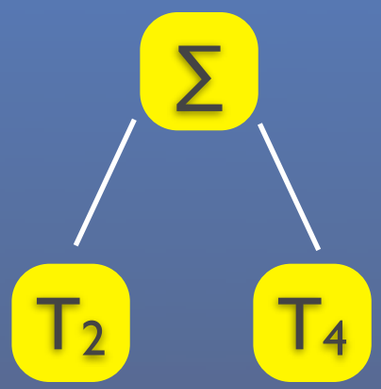


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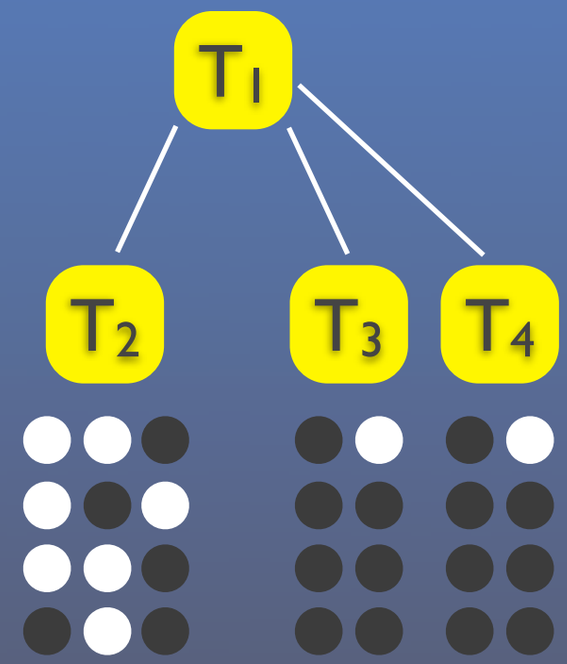


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- Size OTs = size ensembles
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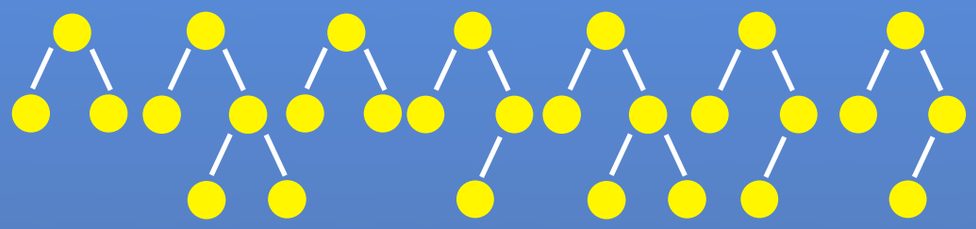

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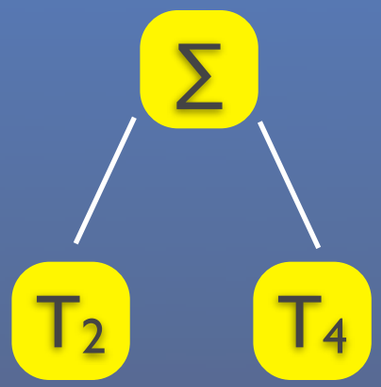


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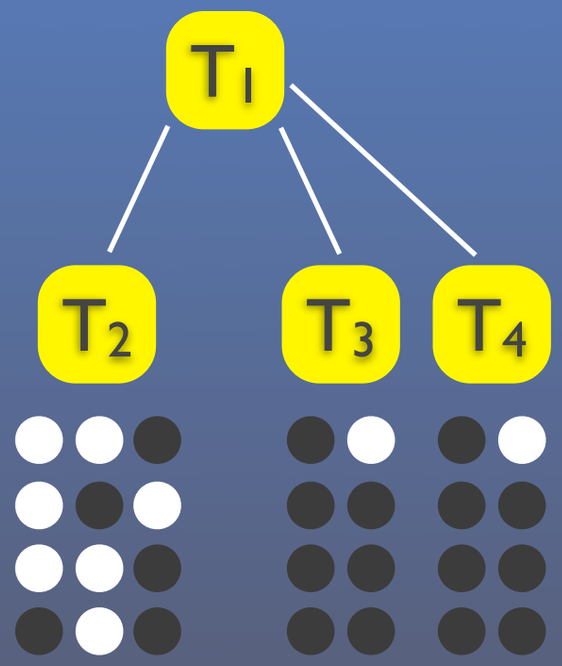


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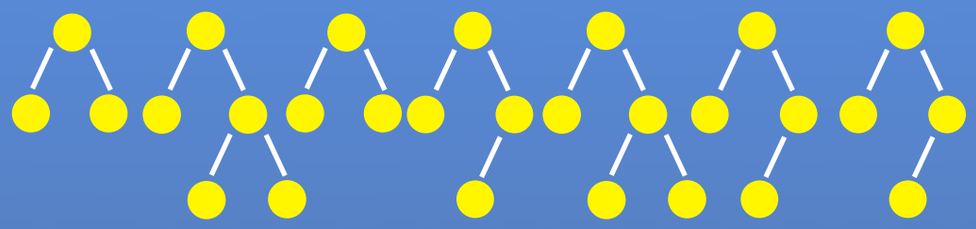

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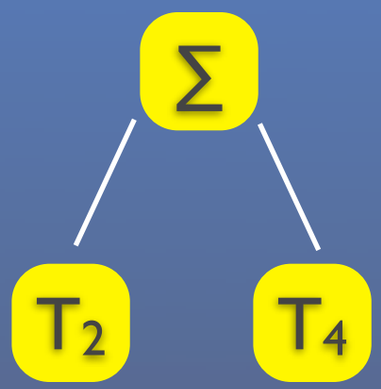


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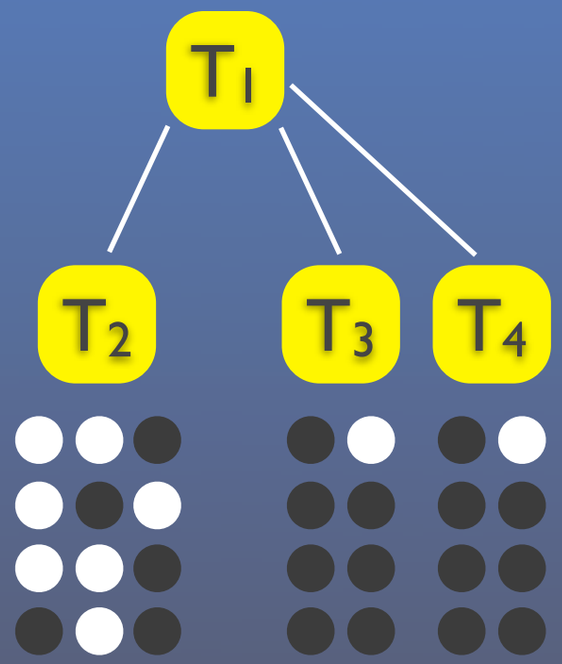


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# Goals

G1: High, consistent accuracy

G2: Small total number of nodes

G3: Structural stability w.r.t. data changes

# Our Solution

Accuracy & stability:

- Ensembles as tests
- Statistically quantified patterns

Size/Interpretability:

- Keep ensembles small
- Conjunctive patterns

# Algorithmic aspects

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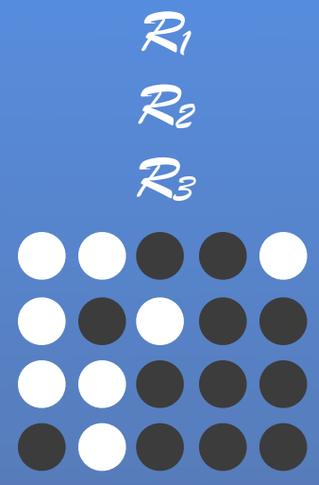
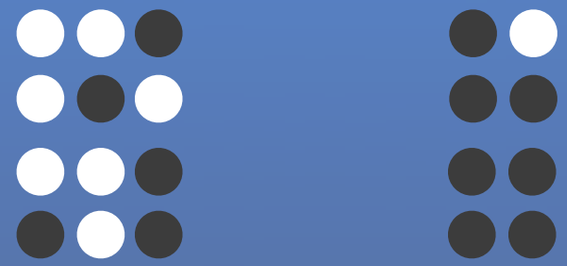
1. Induce *k* best patterns (IG)



# Algorithmic aspects

1. Induce *k* best patterns (IG)
2. Split data

$R_1$
$R_2$
$R_3$



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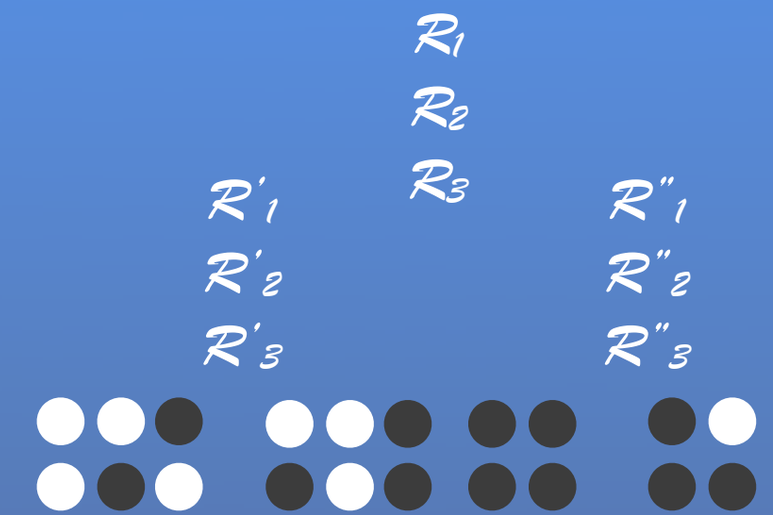
1. Induce *k* best patterns (IG)
2. Split data
3. Repeat until




- Subset too small
- Less than *k* rules found

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Branch-and-bound search:

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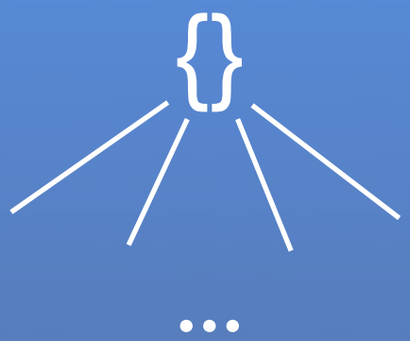
- Enumerate/evaluate pattern(s)



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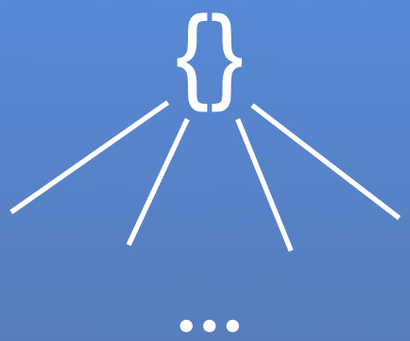


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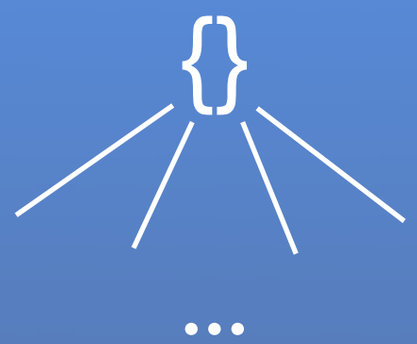


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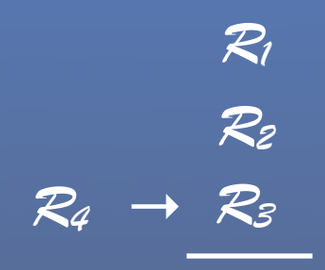

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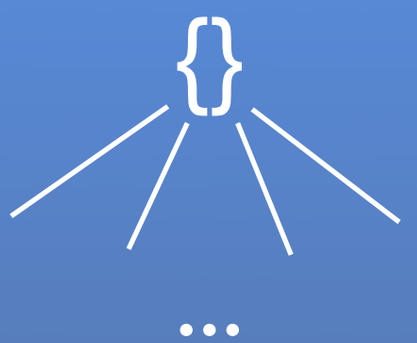


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Branch-and-bound search:

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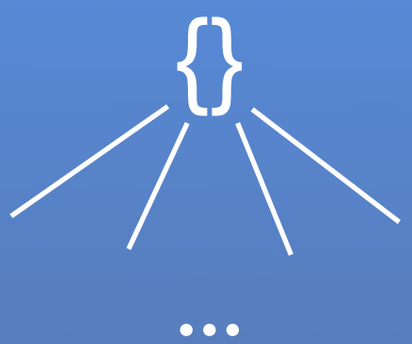


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Branch-and-bound search:

- Enumerate/evaluate pattern(s)


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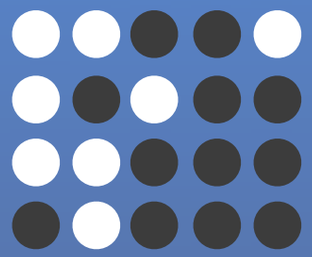
- Prune using upper bounds

# Splitting Data

## One Pattern: simple, match/non-match



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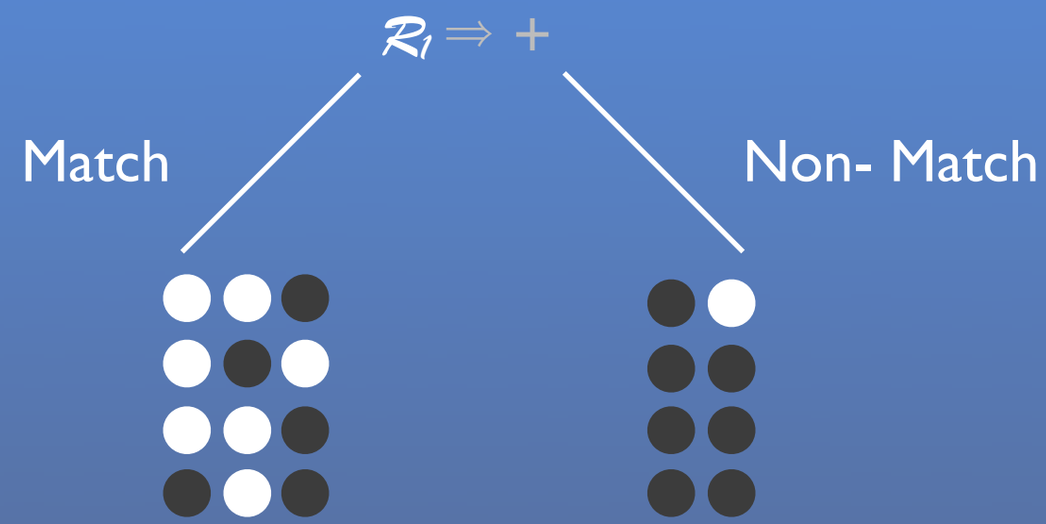

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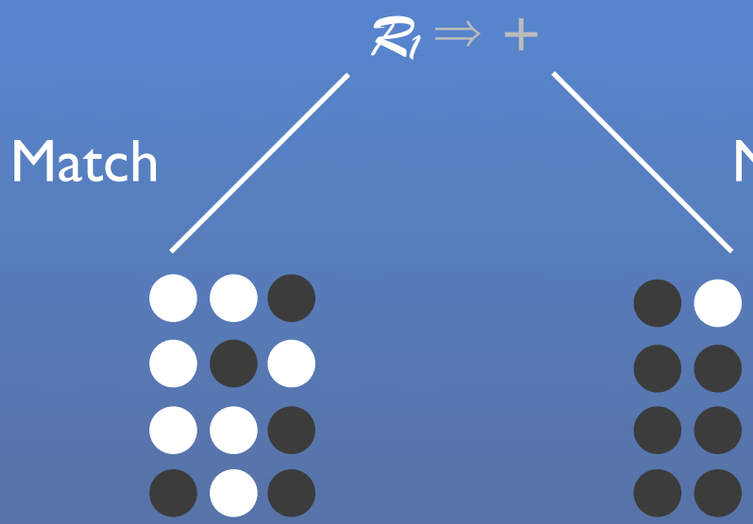
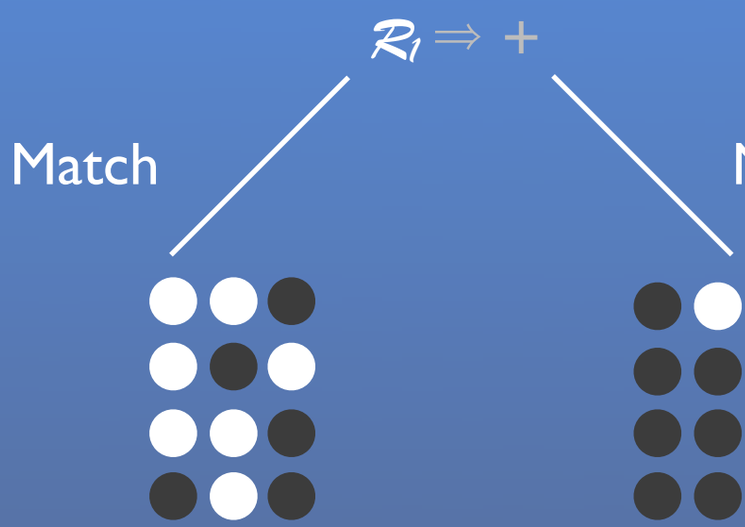
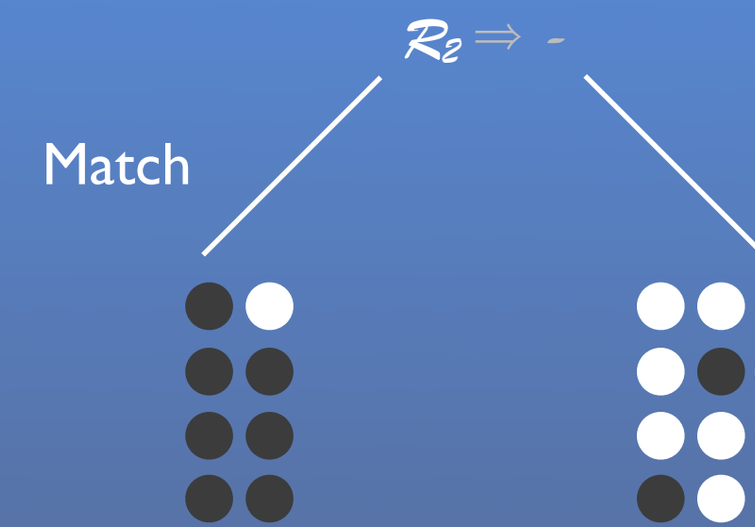


# Splitting Data

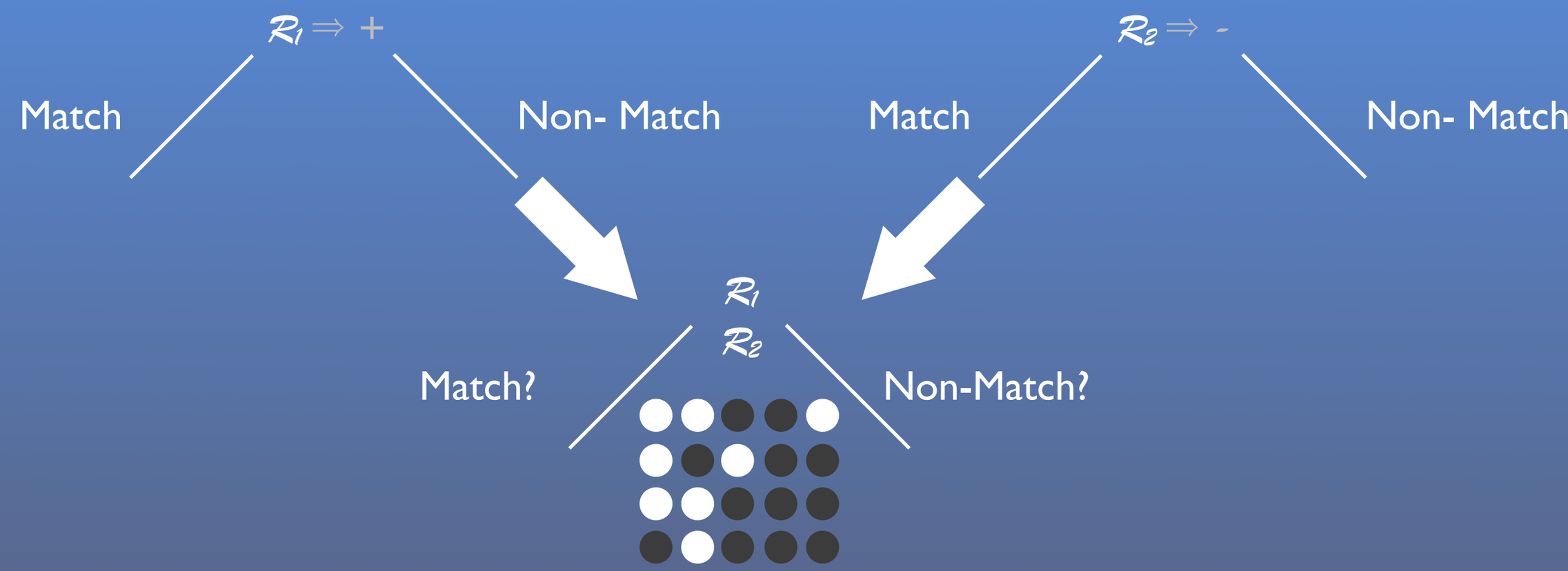
## Several Patterns: less simple



# Splitting Data

## Several Patterns: less simple



# Splitting Data

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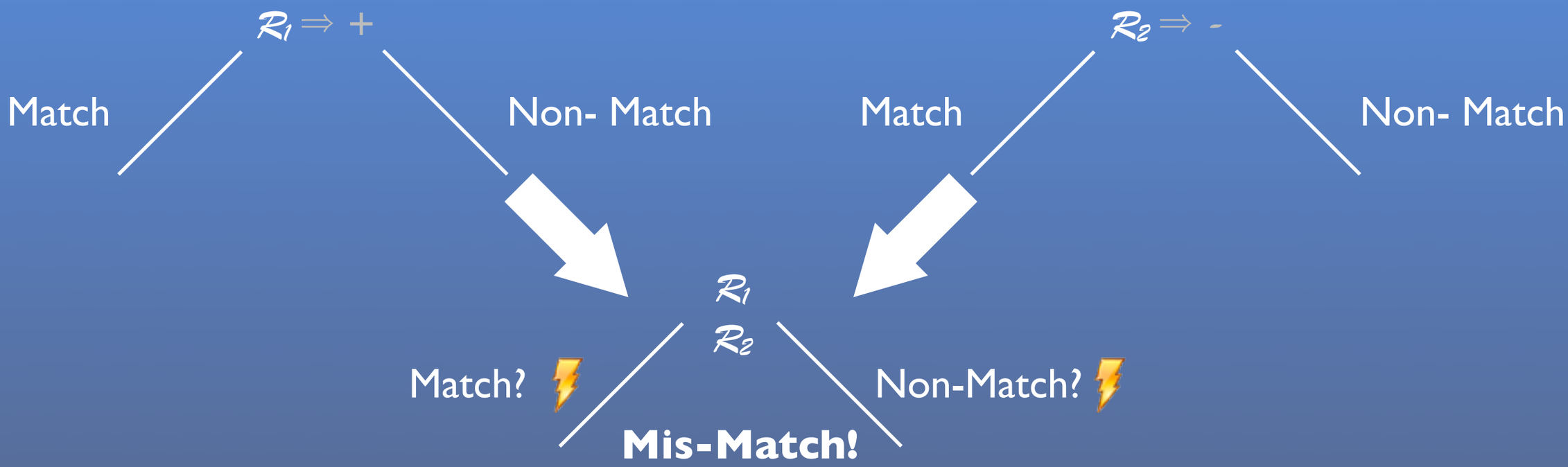
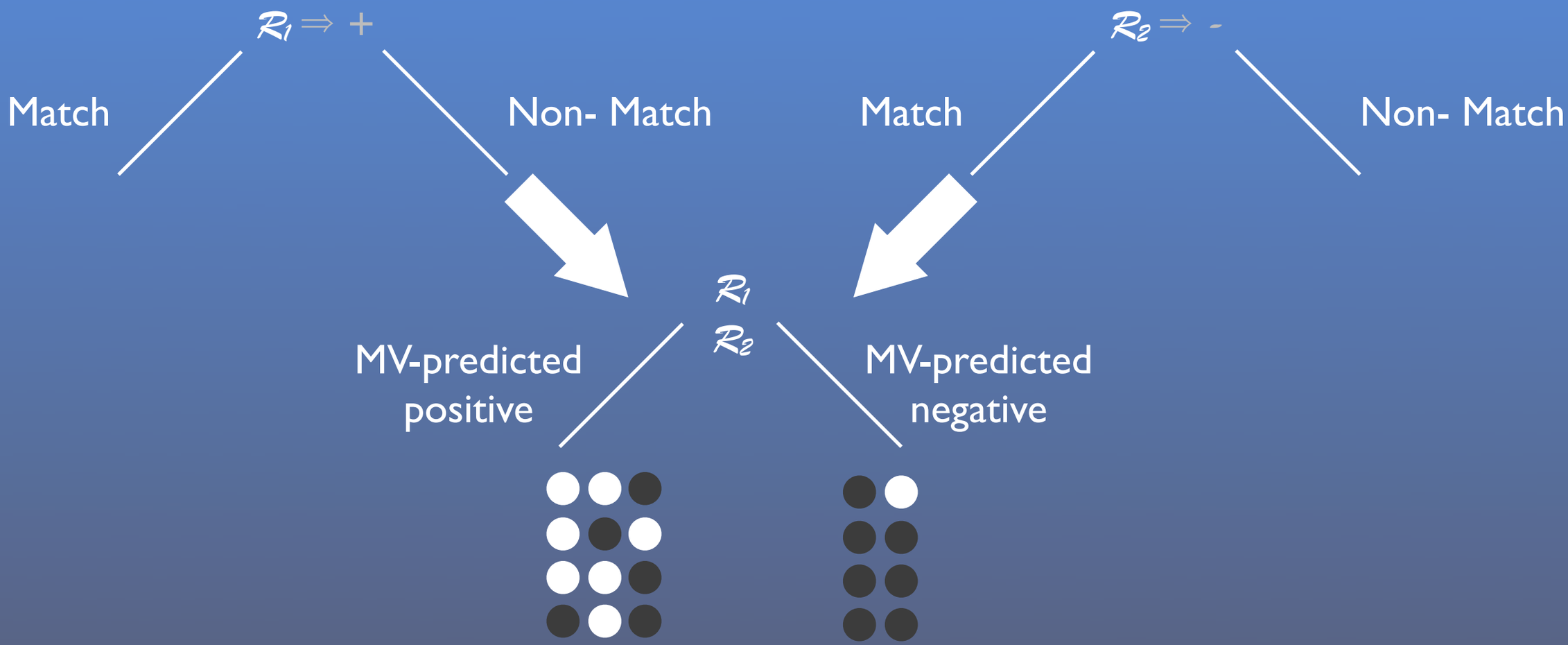


# Splitting Data

## Several Patterns: less simple



# So How Does One Look Like?

$$A_1 = v_2 \wedge A_5 = v_1$$
$$A_3 = v_1 \wedge A_2 = v_1 \wedge A_4 = v_1$$
$$A_3 = v_1$$

} Majority Vote

$$A_1 = v_1$$
$$A_3 = v_1 \wedge A_5 = v_2$$
$$A_4 = v_1 \wedge A_5 = v_2$$

} Majority Vote

+

+

-

# Performance

- Compared to Bagging, Boosting, C4.5
- Good accuracy 
- Lot smaller than ensembles 
  - 1-2 orders of magnitude (# nodes)
- Still as unstable as C4.5, depending on data 

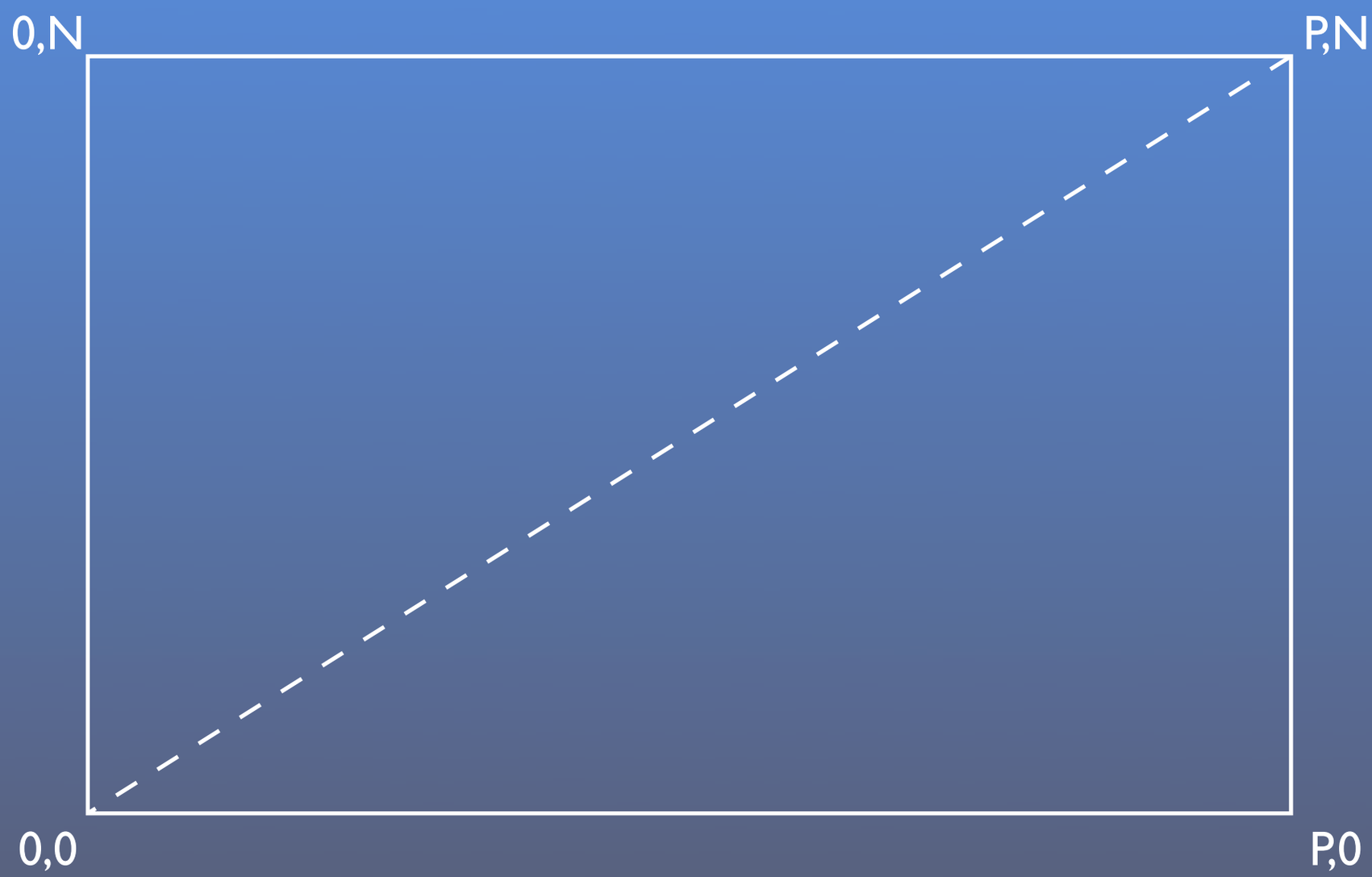
# Summary

- Goal: improve accuracy, stability of DTs
- Invert ensemble process ⇒ inside nodes
- Simple to induce, effective pre-pruning
- Improvements necessary

# But wait...

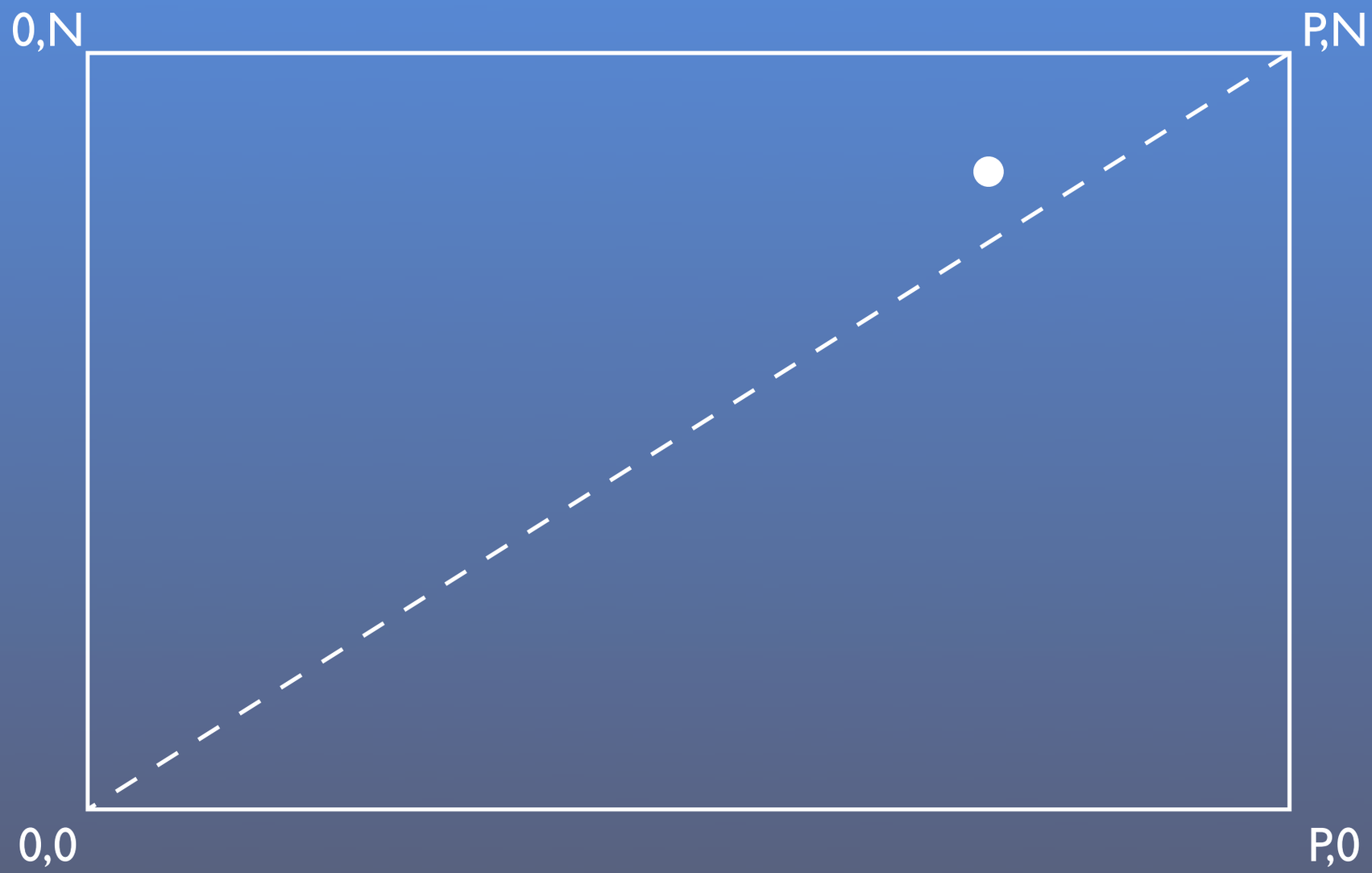
- There's more
- Not limited to classification
- DTs manipulate data

# Manipulating Data (DTs)



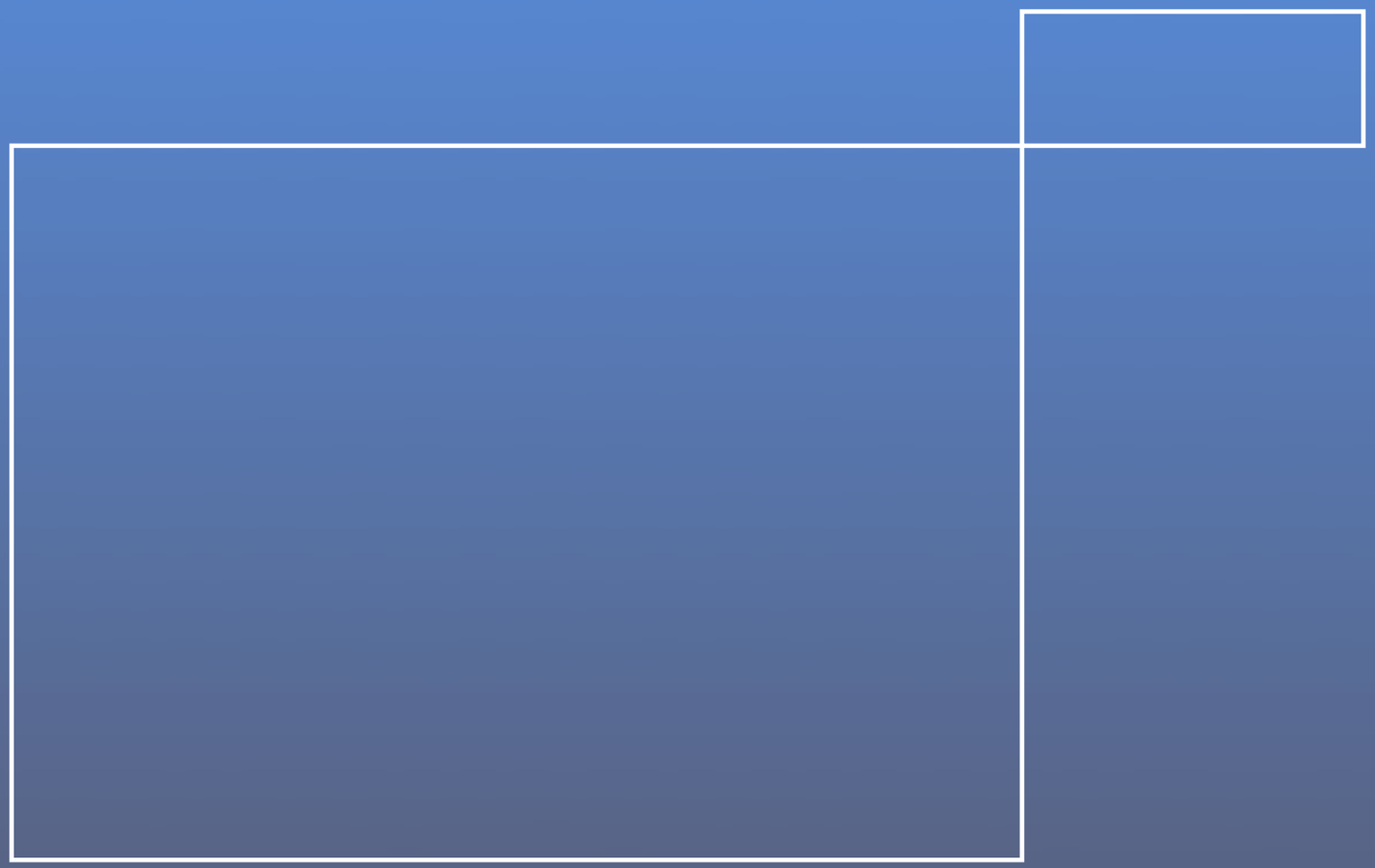
# Manipulating Data (DTs)

$$A_1 = v$$



# Manipulating Data (DTs)

$$A_1 = v$$



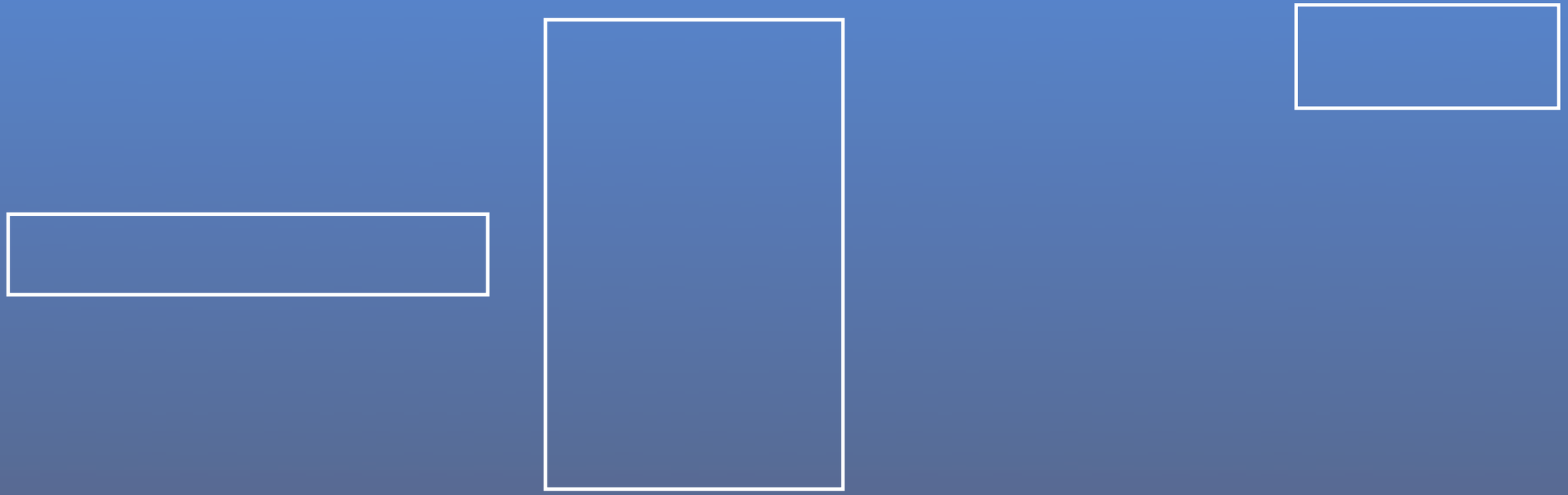
# Manipulating Data (DTs)

$$A_1 = v$$

# Manipulating Data (DTs)

$A_1 = v$

$A_2 = v$

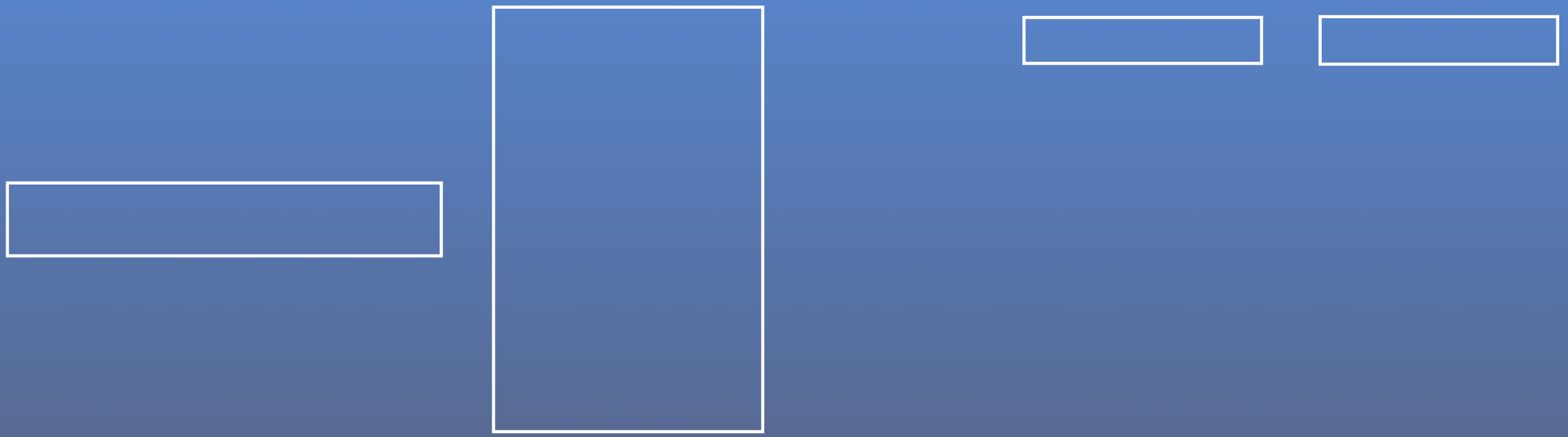


# Manipulating Data (DTs)

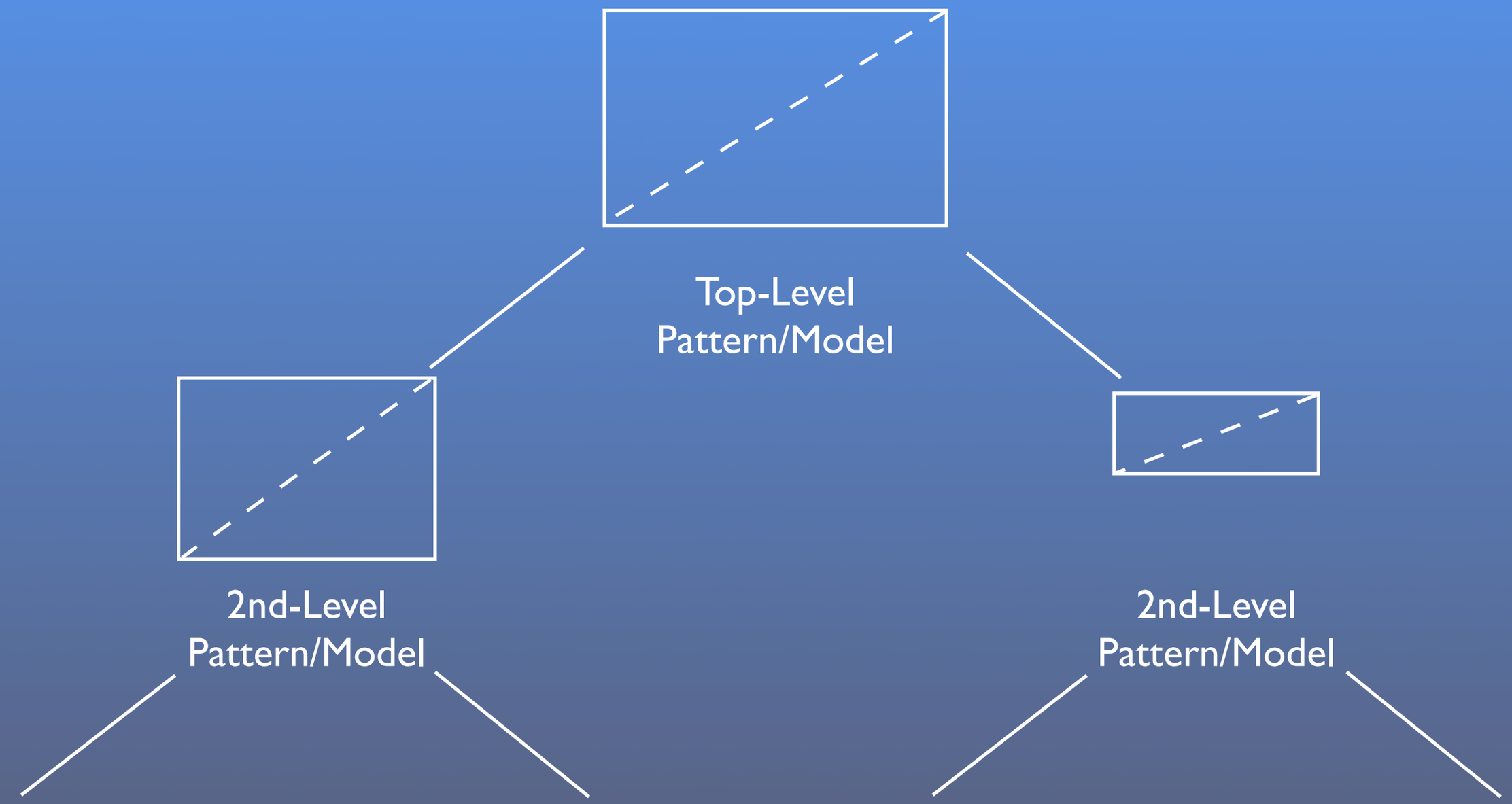
$A_1 = v$

$A_2 = v$

$A_3 = v$

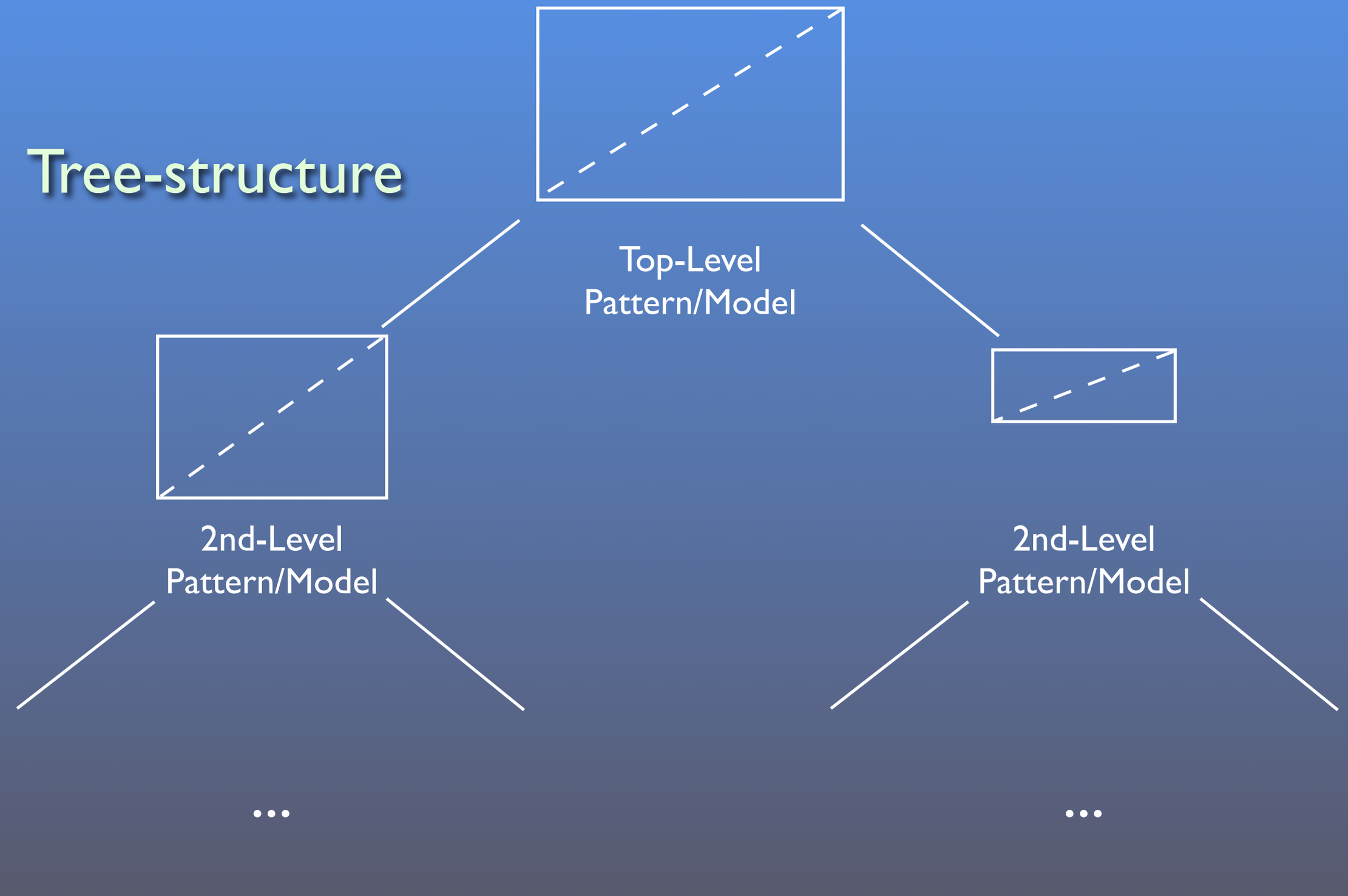


# Tree-Ensembles



# Tree-Ensembles

Tree-structure



# Exist already

- Not flat ensembles (sets)
- Difference in granularity/refinement
- Examples:

Mining interesting, non-redundant patterns:

Tree$^2$, DT-GBI

Building clustering trees (Dendrograms):

CobWeb, TIC, CG-Clus

Using ensembles of conjunctive rules for classification:

Ensemble Trees

# Conclusion

- Goal: improve accuracy, stability of DTs
- Invert ensemble process ⇒ inside nodes
- Simple to induce, effective pre-pruning
- Improvements necessary

# Conclusion

- Goal: improve accuracy, stability of DTs
- Invert ensemble process ⇒ inside nodes
- Simple to induce, effective pre-pruning
- Improvements necessary
- Instantiation of more general mechanism

Rebuttals?

Support?

Questions?

Doubt?