

Inferential Approach to Mining Surprising Patterns in Hypergraphs

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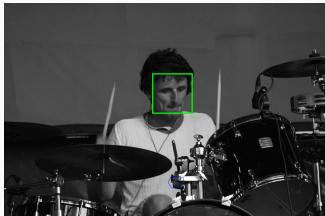
AGI-19, Shenzhen



SingularityNET



Reframing **learning** as reasoning



\Rightarrow

$\mathcal{T} \vdash \mathcal{F}$

Reframing mining surprising patterns as reasoning

1. Learning frequent patterns
2. Assessing their surprisingness

Learning how to **reason efficiently**.

- Unified Rule Engine
 - Evolves Inference Trees TODO: add pic
 - **Control Rules** to select premises and rules

Inference Control Meta-learning

Learning how to **reason efficiently**.

- Unified Rule Engine
 - Evolves Inference Trees TODO: add pic
 - **Control Rules** to select premises and rules
- Learn Control Rules for efficient reasoning TODO: diagram with learning control rules controlling inference.

Mining Frequent Patterns

Brute force algorithm:

- S : minimum support
 - P, Q : patterns
 - \mathcal{C} : pattern pool
 - \mathcal{D} : database
1. Select P from \mathcal{C}
 2. Select *specialization* Q of P such that $S \leq \text{support}(Q, \mathcal{D})$
 3. Add Q to \mathcal{C}
 4. Repeat

Mining Frequent Patterns as Reasoning

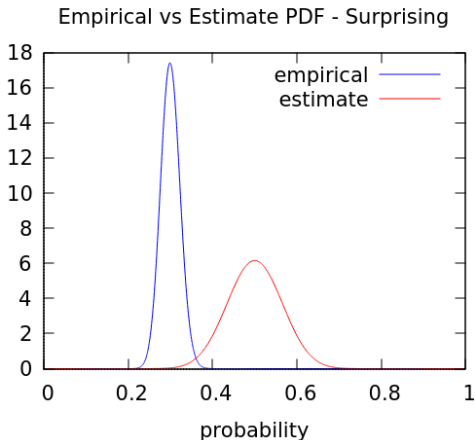
$$\frac{S \leq \text{support}(Q, \mathcal{D}) \quad \text{spec}(Q, P)}{S \leq \text{support}(P, \mathcal{D})} \text{ (AP)}$$

TODO: make mini inference tree expansion example.

Mining Surprising Patterns

Definition

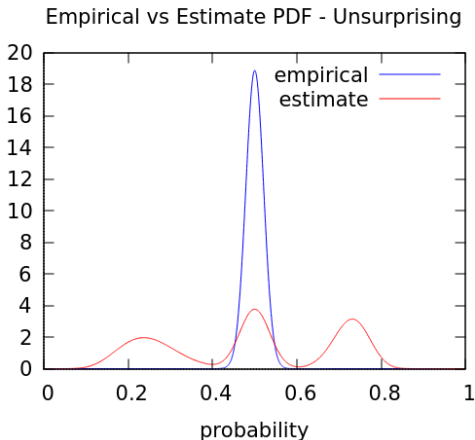
surprise: **contrary to expectation**



Mining Surprising Patterns

Definition

surprise: **contrary to expectation**



Mining Surprising Patterns as Reasoning

TODO

Examples