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Learning Constraint Networks over Unknown Constraint Languages

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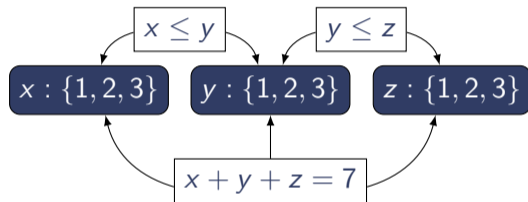
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Background

A **constraint network**:

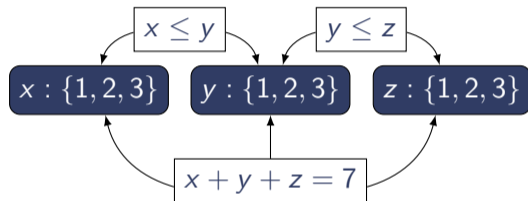
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- a set of **constraints**, i.e. relations between variables that must be satisfied in any **solution**



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A **constraint language** is a set of relations over a domain.

- ▶ Language of a constraint network : set of all relations that appear in its constraints

Constraint Programming & Constraint Acquisition

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Challenge | Designing a constraint network representing a given problem can be difficult.

- ▶ To overcome this, **constraint acquisition** learns a constraint network automatically.

Definition | Passive Constraint Acquisition

Instance: Set of examples, labelled as solutions and non-solutions.

Goal: Find a constraint network that correctly classifies the examples.



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CONACQ.1 (Bessiere et al., 2006, 2017), MODELSEEKER (Beldiceanu and Simonis, 2012),
BAYESACQ (Prestwich et al., 2021), COUNT-CP (Kumar et al., 2022)

Problem | All current approaches require some knowledge of the constraint language of the output network.



Our contribution

We develop a constraint acquisition method that **constructs a constraint language as part of the learning process.**



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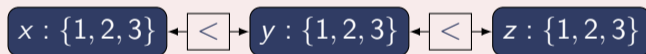
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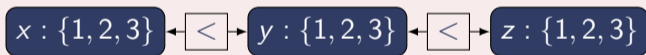
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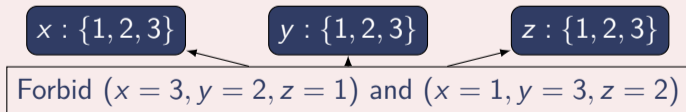
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▶ A bad network:



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Problem | LANGUAGE-FREE ACQ

Instance: Set of examples, labelled as solutions and non-solutions, two integers k and r .

Question: Is there a constraint network **over a language of size at most k and arity at most r** that correctly classifies the examples?

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LANGUAGE-FREE ACQ is NP-complete even for $(k, r) = (1, 1)$.

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- ▶ **Tie-breaking:** lower arity, more constraints, tighter constraints
- ▶ **Construct and solve a model for each (k, r) :**
 - ▶ Convert to an instance WEIGHTED PARTIAL MAX-SAT
 - ▶ Compute the optimal network or prove that none exists
 - ▶ Output the first constraint network found

Experimental Results

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- Learn an **equivalent network** on another language:
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 - ▶ Golomb Ruler (with 10 variables) [3200 examples]
- Neither learn the target language nor an equivalent network:
 - ▶ 8-Queens (coordinates model)

Recap and Future Work

We proposed a novel constraint acquisition method that **does not require any knowledge on the constraint language of the target network.**



Experiments show promising results, but has limitations. Could be addressed by using **more sophisticated notions of simplicity** and **detecting topological information.**



Thank you for your time and attention.



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Experiments performed with the MESO@LR-Platform at University of Montpellier

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