

A Comparison of ASP-Based and SAT-Based Algorithms for the Contension Inconsistency Measure

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Hence, the handling of inconsistent information is a crucial problem.

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- ▶ Complexity study by Thimm and Wallner (2019):
 - ▶ Inconsistency measurement is computationally hard in general
 - ▶ The most suitable candidates for practical applications are on the first level of the polynomial hierarchy
 - ▶ In this work, we consider the *contension inconsistency measure*

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- ▶ We propose a revised version of an algorithm based on *answer set programming* (ASP)
- ▶ **Experimental evaluation:** We compare the two methods to each other, and to a naive baseline method

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- 2 Algorithm Based on SAT
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Definition

Let \mathbb{K} be the set of all (propositional) knowledge bases.

An *inconsistency measure* \mathcal{I} is a function $\mathcal{I} : \mathbb{K} \rightarrow \mathbb{R}_{\geq 0}^{\infty}$ that satisfies $\mathcal{I}(\mathcal{K}) = 0$ iff \mathcal{K} is consistent, for all $\mathcal{K} \in \mathbb{K}$.

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- ▶ A three-valued *model* is an interpretation where each formula $\alpha \in \mathcal{K}$ is assigned either t or b .

Definition

The *set of models* wrt. \mathcal{K} is defined as

$$\text{Models}(\mathcal{K}) = \{\omega^3 \mid \forall \alpha \in \mathcal{K}, \omega^3(\alpha) = t \text{ or } \omega^3(\alpha) = b\}.$$

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$$\text{Conflictbase}(\omega^3) = \{x \in \text{At}(\mathcal{K}) \mid \omega^3(x) = b\}$$

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- ▶ Input formulas must be in *Conjunctive Normal Form* (CNF)
- ▶ *Cardinality constraints* (here: *at-most-k* constraints): $a_1 + \dots + a_n \leq k$
 - ▶ + operator: for every true atom, add 1

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 - ▶ After $\log_2(|\text{At}(\mathcal{K})|)$ calls¹, we know $\text{VALUE}_{\mathcal{I}_c}$

¹ $\text{At}(\mathcal{K})$ refers to the signature size of \mathcal{K}

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- ▶ Cardinality constraint representing that at most u of the b -atoms are true:

$$\text{at_most_}u(\text{At}_b)$$

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- ▶ An extended logic program consists of rules
 - ▶ In addition, we use *cardinality constraints*, and *optimize statements*

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- ▶ There are two previous versions of the ASP-based approach in the literature
- ▶ Our new revision is very similar to the second approach (Kuhlmann and Thimm, 2021), however it uses first-order concepts for ASP rules
 - ▶ This eases readability and
 - ▶ allows for an automated, internally optimized, grounding procedure

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- 3 Algorithm Based on ASP
- 4 Experimental Analysis**
- 5 Conclusion

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- ▶ Synthetic dataset
 - ▶ Created using the *SyntacticRandomSampler*²
 - ▶ 1800 knowledge bases
 - ▶ Smallest instances: signature size 3; 5–15 formulas
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ML Dataset

- ▶ “Translated” dataset
- ▶ Based on *Animals with Attributes*
 - ▶ Using the Apriori algorithm, we mined association rules
 - ▶ Rules were interpreted as propositional logic implications
- ▶ 1920 knowledge bases
- ▶ Mean signature size: 76
- ▶ Mean number of formulas: 11,767

²<http://tweetyproject.org/api/1.14/net/sf/tweety/logics/pl/util/SyntacticRandomSampler.html>

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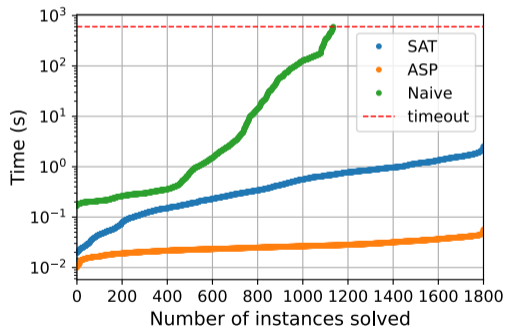
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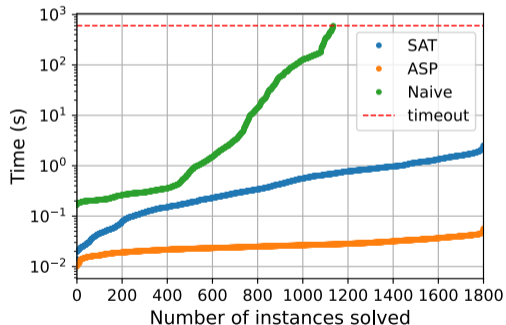
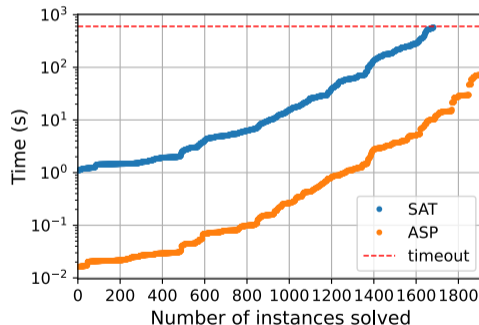
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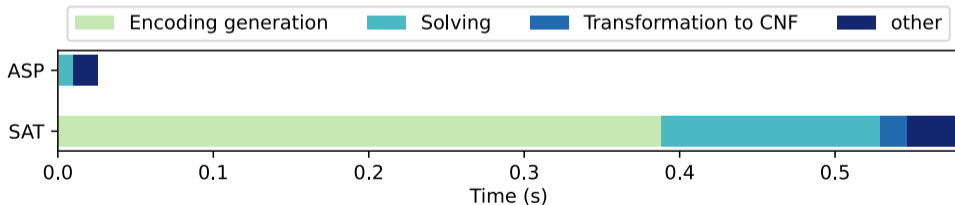
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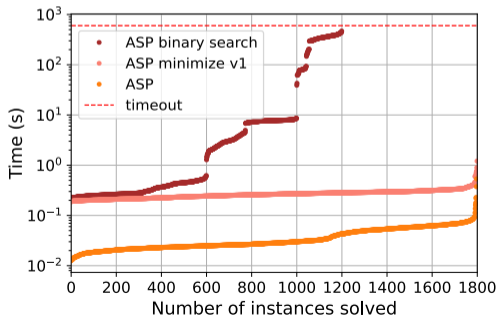
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Thank you for your attention!

We compare the SAT-based and ASP-based approach with a *naive baseline implementation*:

- ▶ The given knowledge base is first converted to CNF and checked for consistency
- ▶ If consistent: return 0
- ▶ Else: for each proposition x , remove each clause containing x and check for consistency again
 - ▶ This is equivalent to setting x to b
- ▶ If one of the new knowledge bases is consistent, return 1
- ▶ Otherwise: repeat the process with each pair of propositions, then with each triple, and so forth