

On Incremental Core-Guided MaxSAT Solving

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Scenario

$$F_1 \subseteq F_2 \subseteq F_3 \subseteq \dots \subseteq F_n$$

$|F_i|$ — up to 10^8

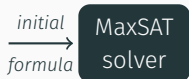
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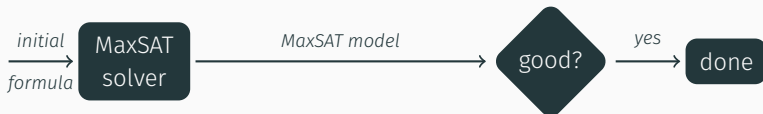


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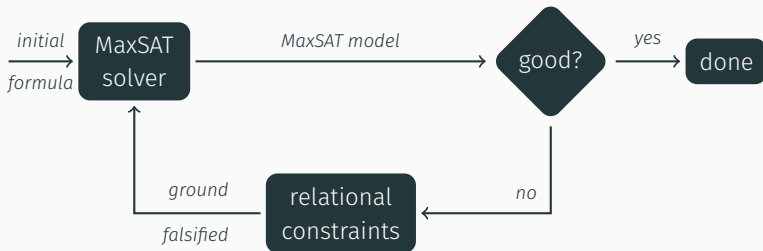
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$(\text{split_lim}_c \leq k \ \forall c \in F_{\text{soft}})$

Experimental results

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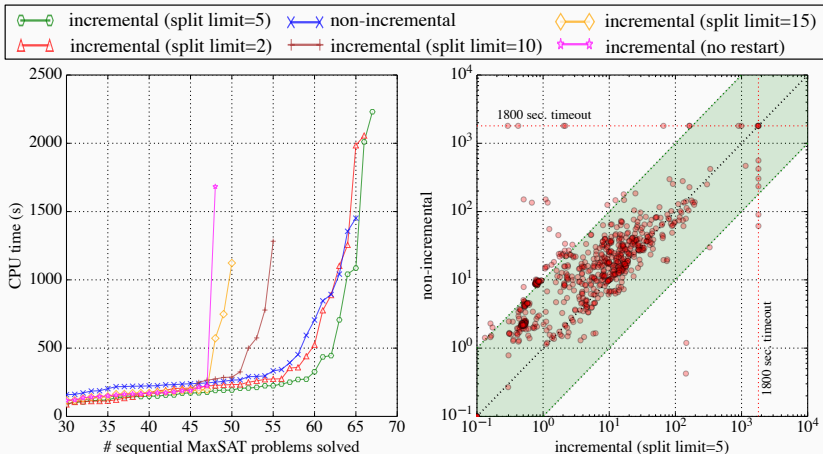
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Experimental results



(a) sequential MaxSAT problems

(b) individual MaxSAT instances

Split limit 5 vs. non-incremental:

- average speedup — $1.8\times$
- best speedup — $296\times$!

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- new **incremental** approach to sequential MaxSAT:
 - incremental MaxSAT calls **+**
 - incremental SAT calls inside MaxSAT **+**
 - adaptive restarts
- better restart strategies
- state-of-the-art MaxSAT algorithms
- not only **add** but also **delete** clauses

Questions?