

Scalable Satisfiability-driven Problem Solving

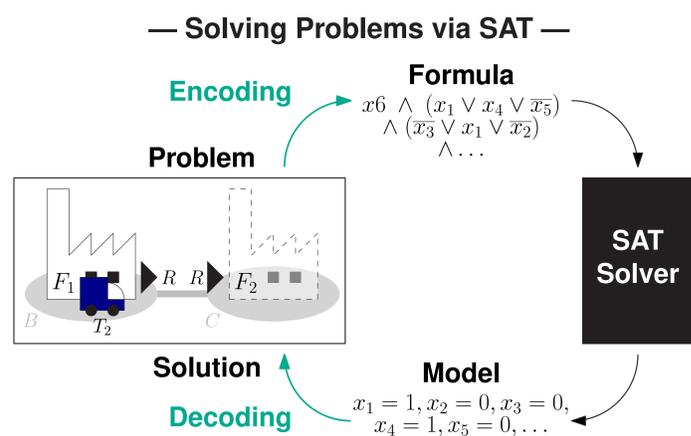
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0. Background

The Propositional Satisfiability Problem

Given a **propositional formula** F , assign a value (true or false) to each of its variables in such a way that F evaluates to true, or report that such an assignment (“model”) does not exist.



Applications: Planning & Scheduling, Explainable AI [1], Theorem Proving, Software & hardware verification, ...

II. A Scalable Encoding for AI Planning [4]

Totally-Ordered Hierarchical Task Network Planning

- Problems feature parametrized **tasks** and their decompositions
- Applications:** Coordination of robots [5], video game AI, web service composition, ...
- Previous SAT-based planners: **Expensive preprocessing** enumerates all possible operations ⇒ **Blowup** in problem size

— Our Approach: Lilotane —

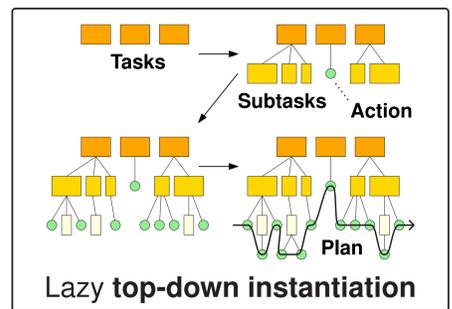
drive T_1 from B to A
drive T_1 from C to A
drive T_2 from B to A
drive T_2 from C to A
...



drive τ from λ to A
 $\tau = T_1 \quad \lambda = B$
 $\vee \tau = T_2 \quad \vee \lambda = C$
 $\vee \dots \quad \vee \dots$



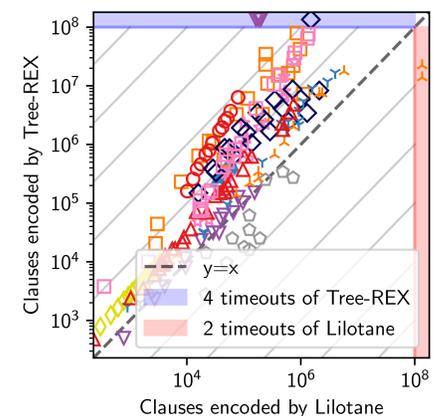
Lifted SAT encoding



SAT-based plan improvement

— Results —

- Smaller formulae** than [6, 7]
- Reduced memory footprint
- Merits of grounding rarely outweigh its problems (see “Entertainment”)
- Competitive:** 2nd place in Total Order track of International Planning Competition 2020
- Produces **high quality plans** even before plan improvement

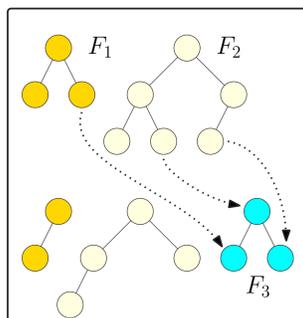
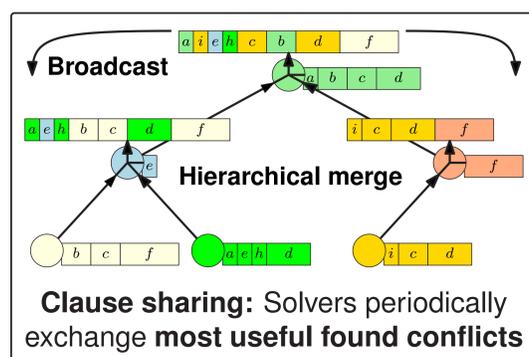


I. Scalable Satisfiability Solving [2]

- Exploit **large clusters** (> 1000 cores)
- Substantially **reduce scheduling latencies** for interactive solving

— Our Approach: Mallob —

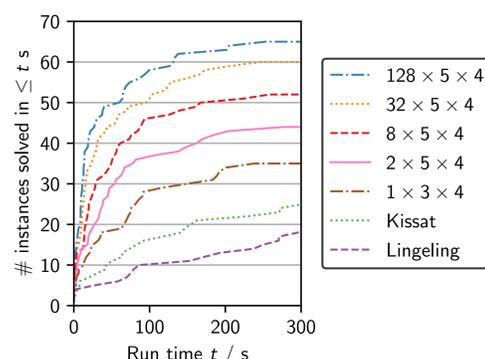
Portfolio approach: Employ many different SAT solvers in parallel [3]



Malleability:
Flexible processing of many formulae

— Results —

- Speedups observed for **up to 2560 cores**
- Winner of International SAT Competitions' Cloud Tracks 2020, 2021
- Malleable scheduling:** Minimal latencies (≤ 10 ms), improved **resource efficiency**



— References —

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