

Accelerating SAT Based Planning with Incremental SAT Solving

1. What is Planning?

- World state: instantiation of multi valued state variables
- Actions:
- require certain values of state variables to be used
- change values of state variables by their effects
- Objective:
- Given a set of actions
- Given an initial state (start) and goal conditions
- Find a plan (sequence of actions to get from start to goal)

3. Finding Plans with Satisfiability Solvers

- Create F_{k} which is satisfiable only if plan of size k exists
- Many encoding schemes for F_{ν} exist
- → use abstract view, such that all can be used:
- initial clauses I: satisfied in the initial state
- goal clauses G:
- satisfied in the goal state transition clauses T:
- satisfied at each pair of consecutive states
- Solve F_1 , F_2 , ... until a satisfiable formula F_n is reached
- Use the solution of F_n to construct a plan

Goal to Init

or Init to Goal?

madagascar 250 non incr. \exists 200 incremental ∃ 150 100

60

Cactusplot comparing madagascar with our approach.

70

80

number of instances solved

90

100

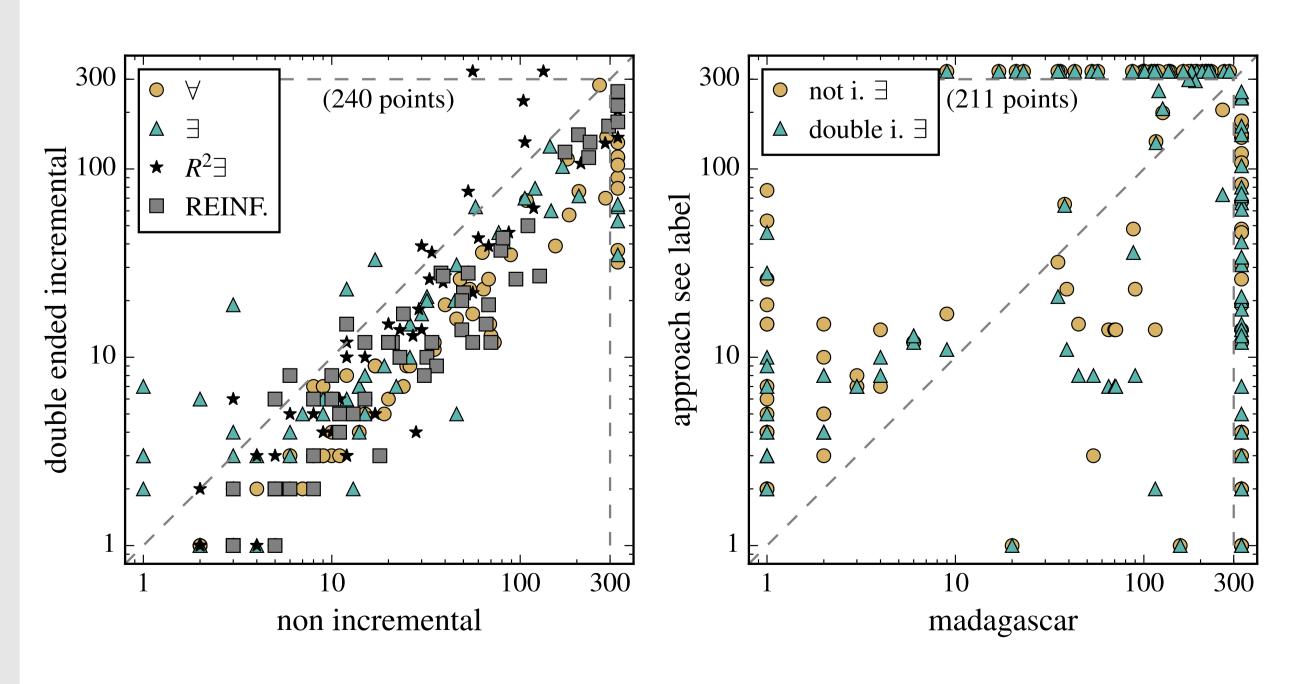
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4. Incremental SAT Solving

- Standardized interface IPASIR
- → profit from future SAT solver development
- Idea: Solve multiple slightly different formulas
- Advantage: Reuse information from previous solve:
- Internally learned clauses
- Importance of Variables and Clauses
- Good assignments (phase saving)

Realization:

- Keep all added Clauses
- Solve with assumptions (temporary unit clauses)
- Allows de-/ activation of clauses via activation literals



The Scatter Plots show a data point for each problem. If a problem is only solved by one approach but not the other it is plotted behind the 300 second.

7. Conclusion

- Incremental SAT solving is very beneficial
- It is important to keep track of SAT solver development

Package locations P and Q $dom(P) = dom(Q) = \{LA, SF, LV, Tr\}$

Los Angeles

Initial State: T=LA, P=LA, Q=SF **Goal Conditions**: P=LV, Q=LV

San Francisco

State Variables and their domains:

2. Example: delivering 2 packages to Las Vegas

Actions: Truck location T, dom(T) = {LA, SF, LV} move(x,y)=[prec: {T=x}, eff: {T=y}]

Los Angeles

- loadP(x)=[prec: {T=x, P=x}, eff: {P=Tr}]
- loadQ(x)=[prec: {T=x, Q=x}, eff: {Q=Tr}]

San Francisco

Las Vegas

- dropP(x)=[prec: {T=x, P=Tr}, eff: {P=x}]
- dropQ(x)=[prec: {T=x, Q=Tr}, eff: {Q=x}] Where x,y are LA, SF, and LV

Plan: loadP(LA), move(LA,SF), loadQ(SF), move(SF,LV), dropP(LV), dropQ(LV)

Las Vegas

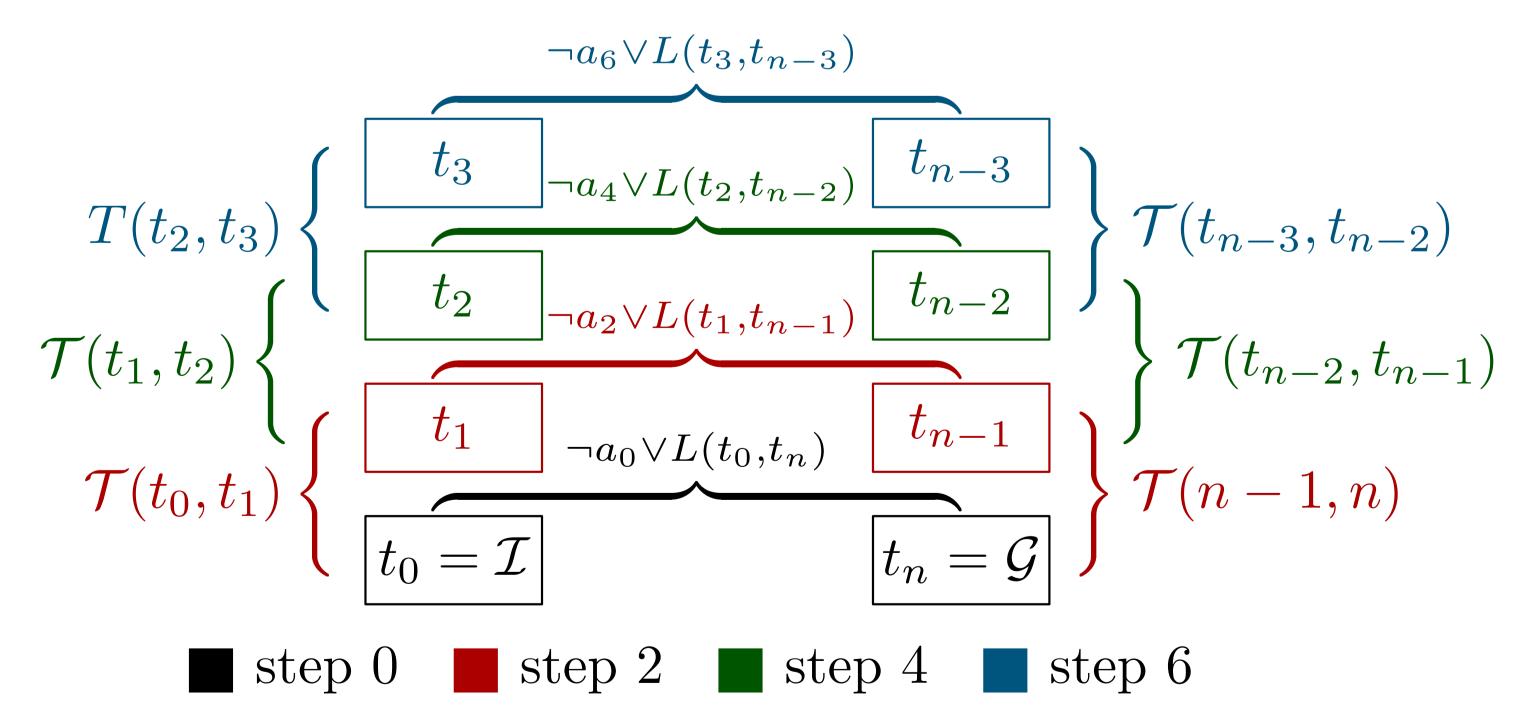
5. Extending F_n to F_{n+1}

Add new timepoints with transitions alternating to goal and initial state.

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- At step n the formula F_n is solved with assumption a_n and contains all clauses of previous steps.
- $L(t_i,t_k)$ is a link clauses such that all variables in t_i have the same value as in t_k and vice versa. As it has an activation literal, only the latest link clause is active.



Double Ended Incremental Encoding up to step 6 and with step size 2.

6. Experiments

- Recent SAT solver from 2016 SAT Competition:
 - COMiniSatPS 2Sun nopre
- Benchmarks:
- Agile Track of the 2014 International Planning Competition (IPC)
- Limits:
- Time limit: 300s, 1 CPU core @ 2.10GHz, 8 GB RAM