

Exploiting optimizations in SAT-based planning: minimal-#actions plans and “soft” goals

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SAT-based planning is the best approach for “optimally” solve planning problems by

- 1 constructing a SAT formula ϕ_n for a fixed *makespan* n ;
- 2 verifying if ϕ_n is satisfiable; if not, n is increased.

Main advantages:

- simplicity
- effectiveness, can take advantage on the continuous progress in the SAT area
- optimal makespan guaranteed

International Planning Competitions

SATPLAN has been the winner of the IPC-4 and co-winner of the IPC-5 in the optimal tracks.

SATPLAN's deficiency

- it can only handle a very limited part of the PDDL language; and
- it does not take into account other “plan quality” issues, e.g., number of actions in the plan and the possibility to express “soft” goals.

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We present SATPLANP, modification of SATPLAN, which returns plans

- 1 having minimal number of actions;
- 2 having maximal number of “soft” goals satisfied.

w.r.t. both

- subset inclusions (qualitative)
- cardinality (quantitative)

Planning as Satisfiability

Planning problem

Is a triple $\langle I, tr, G \rangle$ where (given the sets of fluents \mathcal{F} and actions \mathcal{A})

- I is a SAT formula over \mathcal{F} and represents the set of *initial states*;
- tr is a SAT formula over $\mathcal{F} \cup \mathcal{A} \cup \mathcal{F}'$ where $\mathcal{F}' = \{f' : f \in \mathcal{F}\}$ is a copy of the fluent signature and represents the *transition relation*
- G is a SAT formula over \mathcal{F} and represents the set of *goal states*.

Plan

The *planning problem* Π with *makespan* n is the SAT formula Π_n

$$I_0 \wedge \bigwedge_{i=1}^n tr_i \wedge G_n \quad (n \geq 0) \quad (1)$$

- tr_i is the formula obtained from tr by substituting each symbol $p \in \mathcal{F} \cup \mathcal{A}$ with p_{i-1} and each $f \in \mathcal{F}'$ with f_i
- I_0 and G_n are obvious

A *plan* for Π_n is an interpretation satisfying (1).

SATPLAN's algorithm

function SATPLAN(Π_n)

1 **return** DLL(*cnf*(Π_n), \emptyset)

function DLL(φ, S)

2 **if** ($\emptyset \in \varphi$) **return** FALSE;

3 **if** ($\varphi = \emptyset$) **return** S ;

4 **if** ($\{I\} \in \varphi$) **return** DLL($\varphi_I, S \cup \{I\}$);

5 $I := \text{ChooseLiteral}(\varphi)$;

6 **return** DLL($\varphi_{\bar{I}}, S \cup \{\bar{I}\}$) **or**

7 **return** DLL($\varphi_I, S \cup \{I\}$).

φ_I

φ_I returns the formula obtained from φ by (i) deleting the clauses containing I , and (ii) deleting \bar{I} from the others.

Qualitative SATPLANP's algorithm

function QL-SATPLANP(Π_n, P, \prec)

8 **return** OPT-DLL($\text{cnf}(\Pi_n \wedge \bigwedge_{p \in P} (v(p) \equiv p)), \emptyset, v(P), v(\prec)$)

function OPT-DLL(φ, S, P', \prec')

9 **if** ($\emptyset \in \varphi$) **return** FALSE;

10 **if** ($\varphi = \emptyset$) **return** S;

11 **if** ($\{I\} \in \varphi$) **return** OPT-DLL($\varphi_I, S \cup \{I\}, P', \prec'$);

12 $I := \text{ChooseLiteral}(\varphi, S, P', \prec')$;

13 $V := \text{OPT-DLL}(\varphi_I, S \cup \{I\}, P', \prec')$;

14 **if** ($V \neq \text{FALSE}$) **return** V;

15 **return** OPT-DLL($\varphi_{\bar{I}}, S \cup \{\bar{I}\}, P', \prec'$).

where

- for each $p \in P$, $v(p)$ is a newly introduced variable;
- $v(P)$ is the set of new variables, i.e., $\{v(p) : p \in P\}$;
- $v(\prec) = \prec'$ is the partial order on $v(P)$ defined by $v(p) \prec' v(p')$ iff $p \prec p'$;

Qualitative SATPLANP's algorithm: Example

Going at work from home

$$\begin{array}{l} \neg AtWork_0, \\ AtWork_1 \equiv \neg AtWork_0 \equiv (Car_0 \vee Bus_0 \vee Bike_0), \\ AtWork_1, \end{array} \quad (2)$$

If we have two preferences

- 1 $p_1 = (\neg Bike_0 \wedge \neg Bus_0 \wedge \neg Car_0)$
- 2 $p_2 = (\neg Bike_0 \wedge \neg Bus_0)$

with $p_1 \prec p_2$. OPT-DLL on (2) returns the plan corresponding to $\{Car_0\}$ determined while exploring the branch extending $\{\neg v(p_1), v(p_2)\}$.

Quantitative SATPLANP's algorithm

function QT-SATPLANP(Π_n, P, c)

16 **return** OPT-DLL($cnf(\Pi_n \wedge adder(P, c)), \emptyset, b(c), p(c)$)

adder(P, c) is a SAT formula, e.g., (Warners, IPL 1999)

- if $n = \lceil \log_2((\sum_{p \in P} c(p)) + 1) \rceil$, *adder*(P, c) contains n new variables $\{b_{n-1}, \dots, b_0\} = b(c)$; and
- for any plan π satisfying Π_n , there exists a unique interpretation μ to the variables in $\Pi_n \wedge adder(P, c)$ s.t.
 - 1 μ extends π and satisfies $\Pi_n \wedge adder(P, c)$;
 - 2 $\sum_{p \in P: \pi \models p} c(p) = \sum_{i=0}^{n-1} \mu(b_i) \times 2^i$, where $\mu(b_i)$ is 1 if μ assigns b_i to true, and is 0 otherwise.
- $p(c)$ is the partial order $b_{n-1} \prec b_{n-2} \prec \dots \prec b_0$.

Going at work from home

$$\begin{array}{l} \neg AtWork_0, \\ AtWork_1 \equiv \neg AtWork_0 \equiv (Car_0 \vee Bus_0 \vee Bike_0), \\ AtWork_1, \end{array} \quad (3)$$

If we have two preferences

- 1 $p_1 = (\neg Bike_0 \wedge \neg Bus_0 \wedge \neg Car_0)$
- 2 $p_2 = (\neg Bike_0 \wedge \neg Bus_0)$

with $c(p_1) = 2$ and $c(p_2) = 1$, then two bits b_1 and b_0 are sufficient as output of $adder(\{p_1, p_2\}, c)$. OPT-DLL returns the plan corresponding to $\{Car_0\}$ determined while exploring the branch extending $\neg b_1, b_0$.

Rational and utility of SATPLAN's algorithms

Rational

- we preferentially (and in order) split on variables (literals) defining
 - 1 preferences, in the qualitative case
 - 2 the sum of the weights of preferences, in the quantitative case
- for minimizing actions, the preferences are the atoms related to actions and the split is forced to FALSE
- for maximizing soft goals satisfied, the preferences are the soft goals and the split is forced to TRUE

Utility

- if we want that as few as possible actions are executed, then we have to find an assignment in QT-SATPLAN;
- if we want that no redundant sequence of (possibly parallel) actions is executed, then we have to find an assignment in QL-SATPLAN.

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Experimental analysis: Goals

1. Evaluate SATPLANP w.r.t. the state-of-the-art on problems with “SimplePreferences”, like “soft” goals
 - SATPLANP VS. SGPLAN
2. Evaluate the reductions that can be obtained with SATPLANP over SATPLAN
3. Evaluate the computational costs of such reductions
 - SATPLANP VS. SATPLAN
4. Evaluate what kind of *adder()* works better
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Experimental analysis: 1.

	SGPLAN	SATPLAN	SATPLANP(w)	SATPLANP(b)	SATPLANP(s)
pipe	0/0	0/7	0/18	0/18	0/17
pipet	0/0	0/5	0/11	0/11	0/11
sat	0/10	0/4	0/4	0/4	0/4
air	0/23	0/9	0/11	0/11	0/11
phil	29/0	0/29	0/464	0/464	0/464
opt	12/0	0/12	0/90	0/90	0/90
psr	12/157	0/48	0/231	0/231	0/231
dep	2/3	0/4	0/7	0/7	0/7
driv	0/71	0/10	0/54	0/54	0/50
zeno	0/44	0/9	0/24	0/24	0/24
free	0/12	0/3	0/8	0/8	0/8
log	0/51	0/10	0/33	0/33	0/33
block	0/33	0/9	0/12	0/12	0/12
mpr	0/0	0/4	0/4	0/4	1/3
myst	0/0	0/2	0/2	0/2	0/2
path	7/0	0/7	0/21	0/21	0/21
stor	0/30	0/9	0/10	0/10	0/10
TPP	5/14	0/9	0/27	0/27	0/27
truck	3/0	0/3	0/3	0/3	0/3
Total	70/448	0/193	0/1034	0/1034	1/1028

Table: Results on domains coming from IPCs. x/y stands for x time outs or segmentation faults, y soft goals satisfied.

Experimental analysis: 2. and 4.

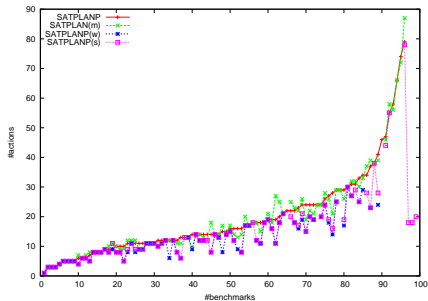
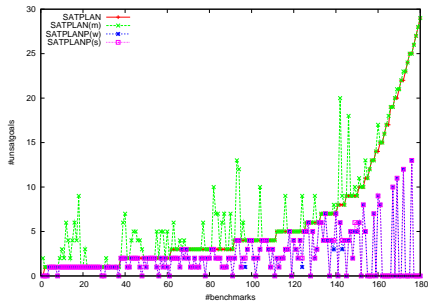


Figure: Left: Number of unsatisfied soft goals by SATPLAN, SATPLAN(m), and SATPLANP(w)/(s). Right: Number of actions in the returned plan for SATPLAN, SATPLAN(m), and SATPLANP(w)/(s).

Experimental analysis: 3. and 4.

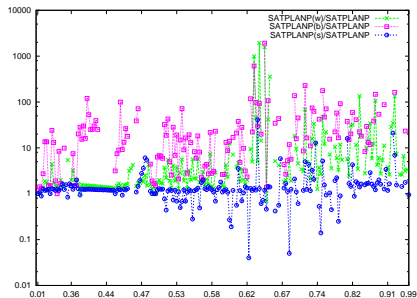
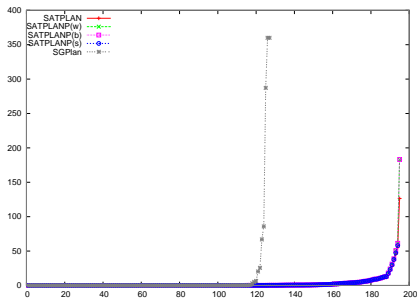


Figure: Left: Performances of SGPLAN, SATPLANP/(w)/(b)/(s). Right: Performances of SATPLANP(w)/(b)/(s) w.r.t. SATPLAN as a function of the ratio between the number of preferences and the number of variables.

Further experiments

PB	#actions					Makespan	
	SATPLAN	SATPLAN(m)	SATPLANP(w)	SATPLANP(s)	SGPLAN	SATPLANP	SGPLAN
log6-0	33	30	25	25	26	9	26
log6-1	28	21	14	16	15	9	15
log6-9	41	39	24	28	28	11	28
block6-0	12	12	12	12	12	12	12
block6-1	10	10	10	10	16	10	16
block6-2	20	20	—	20	32	20	32
stor9	14	14	12	12	11	7	11
stor11	—	—	—	18	17	11	17
stor12	22	25	—	20	17	9	17
sat1	9	9	9	9	9	8	9
sat2	13	—	—	13	13	12	13
psr33	21	21	21	21	21	16	21
psr40	20	20	20	—	20	15	20
psr42	30	30	30	30	30	16	30
driv1	14	18	8	8	7	6	7
zeno5	15	14	14	14	11	5	11
zeno6	14	13	12	12	13	5	13
zeno8	16	17	15	15	12	5	12
free1	9	11	9	11	10	5	10
free2	18	18	—	18	14	8	14
free3	21	21	21	21	19	7	19
air7	41	41	41	41	41	21	41
air9	71	71	—	71	73	27	73
air12	39	39	39	39	39	21	39

Table: #actions and makespan for SATPLAN, SATPLANP and SGPLAN.

Conclusions and future work

Done ... We have

- ✓ extended SAT-based planning to deal with (other) issues related to plan quality;
- ✓ showed that our approach is viable and competitive, often without sacrificing efficiency;
- ✓ evaluated different minimalities and *adder()*s

To be done ...

- ✗ relaxe the optimality in the makespan for further improve the plan quality;
- ✗ experiment with other encodings, other than the “action-based”.

More on this work ...

- SATPLANP's web page at
<http://www.star.dist.unige.it/~marco/SATPLANP/>
- Our ECAI 2006 and AAAI 2007 papers:
 - “Solving Optimization Problems with DLL”; and
 - “Planning as Satisfiability with Preferences”
- Enrico's invited talk at ICAPS'06