## Heuristics for Planning with SAT and Expressive Action Definitions

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## Contribution of the Work

- Earlier work: heuristics for SAT-based planning (classical, non-optimizing), replacing VSIDS et al. in CDCL (Rintanen CP'10, Al'10)
  - A form of backward chaining with CDCL
  - Substantial speed-up for finding plans for most problem types
  - · Applicable to almost all notions of plans used with SAT
- This work extends the heuristic.
  - conditional effects: simple change in the encoding scheme
  - disjunctions: requires bigger changes
- Experimental results: outperforms other planners

## Development of Planning as SAT

(As relevant to the planning techniques in this work)

1992-99	the approach is first developed	Kautz & Selman etc.
2004-06	practical (linear-size) encodings	Rintanen et al.
	no more memory overflows	
2004-06	interleaved search strategies	Rintanen et al.
	efficiency close to best planners	
2010	planning-specific heuristics for SAT	Rintanen
	$efficiency \geq best  planners$	



















Case 1: goal/subgoal x has no support yet

Value of a state variable x at different time points:

	t-8	t-7	t-6	t-5	t-4	t-3	t-2	t-1	t
x	0	0	0		1		1	1	1
action 1	0	0	0			0	0	0	
action 2	0	0		0			0		
action 3	0	0	0	0		0	0		
action 4	0	0			0	0			
		I	I		1	I	I		

Case 1: goal/subgoal x has no support yet

Actions that make x true:

	t-8	t-7	t-6	t-5	t-4	t-3	t-2	t-1	t
x	0	0	0		1		1	1	1
action 1	0	0	0			0	0	0	
action 2	0	0		0			0		
action 3	0	0	0	0		0	0		
action 4	0	0			0	0			
			I	I	I	I	I		

#### The new planning heuristic for CDCL Case 1: goal/subgoal *x* has no support yet

Actions that make x true as early as possible (at t - 5):

	t-8	t-7	t-6	t-5	t-4	t-3	t-2	t-1	t
x	0	0	0		1		1	1	1
action 1	0	0	0			0	0	0	
action 2	0	0		0			0		
action 3	0	0	0	0		0	0		
action 4	0	0			0	0			
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Case 1: goal/subgoal x has no support yet

Choose action 2 or 4 at t-6 as the next CDCL decision variable.

	t-8	t-7	t-6	t-5	t-4	t-3	t-2	t-1	t
x	0	0	0		1		1	1	1
action 1	0	0	0			0	0		
action 2	0	0		0			0		
action 3	0	0	0	0		0	0		
action 4	0	0			0	0			
		I	1				I		

Case 2: goal/subgoal x already has support

Goal/subgoal is already made true at t - 4 by action 4.

	t-8	t-7	t-6	t-5	t-4	t-3	t-2	t-1	t
x	0	0	0		1		1	1	1
action 1	0	0	0			0	0	0	
action 2	0	0		0			0		
action 3	0	0	0	0		0	0		
action 4	0	0		1	0	0			

Use precondition literals of action 4 as new subgoals at t - 5.

Case 2: goal/subgoal x already has support

Goal/subgoal is already made true at t - 4 by action 4.

	t-8	t-7	t-6	t-5	t-4	t-3	t-2	t-1	t
x	0	0	0		1		1	1	1
action 1	0	0	0			0	0	0	
action 2	0	0		0			0		
action 3	0	0	0	0		0	0		
action 4	0	0		1	0	0			

Use precondition literals of action 4 as new subgoals at t - 5.

## Extension to Conditional Effects and Disjunction

- Conditional Effects (without disjunction)
  - Simple change of encoding scheme
- Disjunction
  - Complex subgoals reduced to a set of atomic subgoals.
  - Conceptually a bit more complicated.
  - Need to develop more heuristics for doing this right.

## Encoding for Conditional Effects (No Disjunction)

- Idea: View  $\langle \phi, \{a \rhd b, d \rhd e\} \rangle$  as interdependent STRIPS actions  $\langle \phi \land a, b \rangle$  and  $\langle \phi \land d, e \rangle$ .
- New propositional variable for every conditional effect clause.
- The heuristic uses these variables exactly like action variables.
- Interdependencies of conditional effects handled automatically.

# Example Action $\langle \phi, \{ \underbrace{a \succ b}_{x_1}, \underbrace{d \succ e}_{x_2} \} \rangle$ is translated into

 $o@t \to \phi@t$   $(o@t \land a@t) \leftrightarrow x_1@t$  $x_1@t \to b@(t+1)$ 

 $(o@t \wedge d@t) \leftrightarrow x_2@t$  $x_2@t \to e@(t+1)$ 

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#### Example

Action 
$$\langle \phi, \{\underbrace{a \succ b}_{x_1}, \underbrace{d \succ e}_{x_2} \} \rangle$$
 is translated into

$$\begin{array}{ll}
o@t \to \phi@t \\
(o@t \wedge a@t) \leftrightarrow x_1@t \\
x_1@t \to b@(t+1)
\end{array} \quad (o@t \wedge d@t) \leftrightarrow x_2@t \\
x_2@t \to e@(t+1)
\end{array}$$

#### 3-valued states to evaluate goals in

False: l is false (at t).

True: l is true and there is an explanation for that.

Undetermined: none of the above.

For a goal  $\Phi$ , compute atomic subgoals  $\{l_1, \ldots, l_n\}$  such that

 $\{l_1,\ldots,l_n\}\models \Phi.$ 

- NP-hard to do this.
- NP-hard to minimize n.
- We use an approximation.

#### Disjunctive Goals and Subgoals (preconditions, conditions) (We assume NNF here)

For goal  $\Phi$  and partial state s, compute subg<sub>s</sub> $(\Phi) = \{l_1, \ldots, l_n\}$  (assumption:  $s \models \Phi$ ):

- $\texttt{1} \hspace{0.1 cm} \operatorname{subg}_s(\phi_1 \wedge \phi_1) = \operatorname{subg}_s(\phi_1) \cup \operatorname{subg}_s(\phi_2)$
- ② subg<sub>s</sub>( $\phi_1 \lor \phi_1$ ) = either subg<sub>s</sub>( $\phi_1$ ) or subg<sub>s</sub>( $\phi_2$ ), (one that is true in s.)
- **3** subg<sub>s</sub> $(l) = \{l\}$  if l is Undetermined,  $\emptyset$  otherwise.

This is P-time. Heuristics can be used in case 2.

## Experiments with non-STRIPS IPC Benchmarks

- Generally a runtime improvement is obtained, similarly to IPC STRIPS benchmarks.
- Both M and Mp outperform LAMA, FF. (Other well-known and efficient planners (e.g. YAHSP, LPG) only do STRIPS.)

#### Impact on Runtimes



#### Impact on Plan Sizes



## Comparison by Domain

		Мр	Μ	LAMA11	FF
1998-ASSEMBLY-ADL	24	23	18	24	24
2000-ELEVATOR-FULL	143	138	139	135	132
2000-ELEVATOR-SIMPLE	150	150	150	150	150
2000-SCHEDULE-ADL	150	149	144	150	150
2002-SATELLITE-ADL	20	20	20	20	20
2004-AIRPORT-ADL	50	26	21	33	21
2004-OPTICAL-TELEGRAPH-ADL	48	22	17	2	11
2004-PHILOSOPHERS-ADL	48	48	48	12	12
2006-PATHWAYS-ADL	30	30	30	29	16
2006-TRUCKS-ADL	29	15	14	14	11
2008-OPENSTACKS-ADL	30	15	14	30	30
total	722	636	615	599	577
weighted score	11	8.91	8.41	8.35	7.73

## Comparison by Number of Instances Solved



## Conclusions

What we have done so far:

- We presented variable selection heuristics for planning within the CDCL framework, for general PDDL actions.
- As with STRIPS, this beats other planners by a clear margin.

Future work:

- Combine this with VSIDS to do still better.
- Try with Bounded LTL Model-Checking, Discrete Event Systems diagnosis, ... .