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Lecture slides for Automated Planning: Theory and Practice

### Chapter 3 Complexity of Classical Planning

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### **Review: Classical Representation**

- Function-free first-order language *L*
- Statement of a classical planning problem:  $P = (s_0, g, O)$
- $s_0$ : initial state a set of ground atoms of L
- g: goal formula a set of literals
- Operator: (name, preconditions, effects)

```
take(crane1,loc1,c3,c1,p1)
```

;; crane crane1 at location loc1 takes c3 off c1 in pile p1 precond: belong(crane1,loc1), attached(p1,loc1), empty(crane1), top(c3,p1), on(c3,c1) effects: holding(crane1,c3), ¬empty(crane1), ¬in(c3,p1),

¬top(c3,p1), ¬on(c3,c1), top(c1,p1)

• Classical planning problem:  $\mathcal{P} = (\Sigma, s_0, S_g)$ 

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#### **Review: Set-Theoretic Representation**

• Like classical representation, but restricted to propositional logic

- State: a set of propositions these correspond to ground atoms
  - {on-c1-pallet, on-c1-r1, on-c1-c2, ..., at-r1-l1, at-r1-l2, ...}
- No operators, just actions

take-crane1-loc1-c3-c1-p1 precond: belong-crane1-loc1, attached-p1-loc1, empty-crane1, top-c3-p1, on-c3-c1 delete: empty-crane1, in-c3-p1, top-c3-p1, on-c3-p1 add: holding-crane1-c3, top-c1-p1

Weaker representational power than classical representation
 Problem statement can be exponentially larger

## **Review: State-Variable Representation**

• A state variable is like a record structure in a computer program

Instead of on(c1,c2) we might write cpos(c1)=c2

• Load and unload operators:

```
\begin{array}{l} \mathsf{load}(c,r,l) \\ \texttt{;; robot } r \ \mathsf{loads \ container} \ c \ \mathsf{at \ location} \ l \\ \texttt{precond: } \mathsf{rloc}(r) = l, \mathsf{cpos}(c) = l, \mathsf{rload}(r) = \mathsf{nil} \\ \texttt{effects: } \ \mathsf{rload}(r) \leftarrow c, \mathsf{cpos}(c) \leftarrow r \end{array}
```

```
unload(c, r, l)
;; robot r unloads container c at location l
precond: rloc(r) = l, rload(r) = c
effects: rload(r) \leftarrow nil, cpos(c) \leftarrow l
```

• Equivalent power to classical representation

• Each representation requires a similar amount of space

Each can be translated into the other in low-order polynomial time

- Classical representation is more popular, mainly for historical reasons
  - In many cases, state-variable representation is more convenient

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## **Motivation**

- Recall that in classical planning, even simple problems can have huge search spaces
  - Example:
    - » DWR with five locations, three piles, three robots, 100 containers
    - » 10<sup>277</sup> states



- » About 10<sup>190</sup> times as many states as there are particles in universe
- How difficult is it to solve classical planning problems?
- The answer depends on which representation scheme we use
  - Classical, set-theoretic, state-variable

# Outline

- Background on complexity analysis
- Restrictions (and a few generalizations) of classical planning
- Decidability and undecidability
- Tables of complexity results
  - Classical representation
  - Set-theoretic representation
  - State-variable representation

## **Complexity Analysis**

- Complexity analyses are done on *decision problems* or *language-recognition problems* 
  - ◆ A language is a set *L* of strings over some alphabet *A*
  - Recognition procedure for *L* 
    - » A procedure R(x) that returns "yes" iff the string x is in L
    - » If x is not in L, then R(x) may return "no" or may fail to terminate
- Translate classical planning into a language-recognition problem
- Examine the language-recognition problem's complexity

## Planning as a Language-Recognition Problem

• Consider the following two languages:

PLAN-EXISTENCE =  $\{P : P \text{ is the statement of a planning} problem that has a solution}$ 

PLAN-LENGTH = {(P,n) : P is the statement of a planning problem that has a solution of length  $\leq n$ }

- Look at complexity of recognizing PLAN-EXISTENCE and PLAN-LENGTH under different conditions
  - Classical, set-theoretic, and state-variable representations
  - Various restrictions and extensions on the kinds of operators we allow

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## Complexity of Language-Recognition Problems

- Suppose *R* is a recognition procedure for a language *L*
- Complexity of *R* 
  - $T_R(n)$  = worst-case runtime for *R* on strings in *L* of length *n*
  - S<sub>R</sub>(n) = worst-case space requirement for R on strings in L of length n
- Complexity of recognizing L
  - $T_L$  = best asymptotic time complexity of any recognition procedure for L
  - S<sub>L</sub> = best asymptotic space complexity of any recognition procedure for L

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# **Complexity Classes**

#### • Complexity classes:

- NLOGSPACE (nondeterministic procedure, logarithmic space)
   ⊆ P (deterministic procedure, polynomial time)
   ⊆ NP (nondeterministic procedure, polynomial time)
   ⊆ PSPACE (deterministic procedure, polynomial space)
   ⊆ EXPTIME (deterministic procedure, exponential time)
   ⊆ NEXPTIME (nondeterministic procedure, exponential time)
   ⊆ EXPSPACE (deterministic procedure, exponential time)
   ⊆ EXPSPACE (deterministic procedure, exponential time)
- Let *C* be a complexity class and *L* be a language
  - Recognizing L is C-hard if for every language L' in C, L' can be reduced to L in a polynomial amount of time
    - » NP-hard, PSPACE-hard, etc.
  - Recognizing L is C-complete if L is C-hard and L is also in C
    - » NP-complete, PSPACE-complete, etc.

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## **Possible Conditions**

- Do we give the operators as input to the planning algorithm, or fix them in advance?
- Do we allow infinite initial states? +
- Do we allow function symbols?
- Do we allow negative effects?
- Do we allow negative preconditions?
- Do we allow more than one precondition?
- Do we allow operators to have conditional effects?\*
  - i.e., effects that only occur when additional preconditions are true

These take us outside classical planning

## **Decidability of Planning**



 $^{\alpha}$ This is ordinary classical planning.

 $^{\beta}$ True even if we make several restrictions (see text).

Next: analyze complexity for the decidable cases

## **Complexity of Planning**

 $^{\gamma}$  PSPACE-complete or NP-complete for some sets of operators

Kind of	How the	Allow	Allow	Complexity	Complexity	
represen-	operators	negative	negative	of PLAN-	of FLAN-	
tation	are given	effects?	precon-	EXISTENCE	LENGTH	
			ditions?			
		yes	yes/no	EXPSPACE-	NEXPTIME-	
classical				complete	complete	
rep.	in the		yes	NEXPTIME-	NEXPTIME-	
	input			complete	complete	
		no	no	EXPTIME-	NEXPTIME-	
				complete /	complete	
$\alpha$ no operator has >1 precondition $no^{\alpha}$				PSPACE-	PSPACE-	
	-			complete	complete	
		yes	yes/no	PSPACE $\gamma$	PSPACE $\gamma$	
	in		yes	NP $\gamma$	NP $\gamma$	
	advance	no	no	Р	NP $\gamma$	
			$no^{\alpha}$	NLOGSPACE	NP	

#### • **Caveat:** these are *worst-case* results

- Individual planning domains can be much easier
- Example: both DWR and Blocks World fit here, but neither is that hard
  - For them, PLAN-EXISTENCE is in P and PLAN-LENGTH is NP-complete

Kind of	How the	Allow	Allow	Complexity	Complexity	
represen-	operators	negative	negative	of plan-	of plan-	
tation	are given	effects?	precon-	EXISTENCE	LENGTH	
			ditions?			
		yes	yes/no	EXPSPACE-	NEXPTIME-	
classical				complete	complete	
rep.	in the		yes	NEXPTIME-	NEXPTIME-	
	input			complete	complete	
		no	no	EXPTIME-	NEXPTIME-	
				complete	complete	
			$no^{\alpha}$	PSPACE-	PSPACE-	
			$\downarrow$	complete	complete	
		yes	yes/no	PSPACE $\gamma$	PSPACE $\gamma$	
	in advance		yes	NP $\gamma$	NP $\gamma$	
		no	no	Р	NP $\gamma$	
			$no^{\alpha}$	NLOGSPACE	NP	

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#### • Often PLAN-LENGTH is harder than PLAN-EXISTENCE

#### But it's easier here:

• We can cut off every search path at depth *n* 

Kind of	How the	Allow	Allow	Complexity	Complexity	
represen-	operators	negative	negative	of RLAN-	of plan-	
tation	are given	effects?	precon-	EXISTENCE	LENGTH	
			ditions?			
		yes	yes/no	EXPSPACE-	NEXPTIME-	
classical				complete	$\operatorname{complete}$	
rep.	in the		yes	NEXPTIME-	NEXPTIME-	
	input			complete	complete	
		no	no	EXPTIME-	NEXPTIME-	
				complete	complete	
			$no^{\alpha}$	PSPACE-	PSPACE-	
				complete	complete	
		yes	yes/no	PSPACE $\gamma$	PSPACE $\gamma$	
	in advance		yes	NP $\gamma$	NP $\gamma$	
		no	no	Р	NP $\gamma$	
			$no^{\alpha}$	NLOGSPACE	NP	

# Equivalences

- Set-theoretic representation and ground classical representation are basically identical
  - For both, exponential blowup in the size of the input
  - Thus complexity looks smaller as a function of the input size
- Classical and state-variable representations are equivalent, except that some of the restrictions aren't possible in state-variable representations
  - Hence, fewer lines in the table



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	Kind of	How the	Allow		Allow		Complexity	Complexity	
	represen-	operators	negati	ve	negative		of plan-	of PLAN-	
	tation	are given	effects	?	precon-		EXISTENCE	LENGTH	
				ditions?					
			yes		yes/no		PSPACE-	PSPACE	-
	set-						complete	complet	te
	theoretic	in the			yes		NP-complete	NP-com	plete
	or	input	no		no		Р	NP-complete	
	ground	ground				β	NLOGSPACE- NP-		
	classical						complete	complet	te
	rep.	in	yes/no	> /	yes/no		constant	constan	t
		advance					time	$\operatorname{time}$	
ſ	state-	in the	$yes^{\delta}$		yes/no		EXPSPACE-	NEXPTI	ME-
Like	variable	input					complete	complet	te
classical	rep.	in	$y q s^{\delta}$		yes/no		PSPACE $\gamma$	PSPACE	γ
rep, but ∫		advance							
fewer	ground	in the	$yes^{\delta}$		yes/no		PSPACE-	PSPACE	-
lines in the table	state-	input /					complete	complet	te
	variable	in	$yes^{\delta}$		yes/no		constant	constan	ıt
	rep.	advance					time	time	
$^{\alpha}$ no operator has >1 precondition				$^{\beta}$ every operator with >1 precondition is the composition of other operators $_{17}$					



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