



# Planning and Scheduling: Complexity of Classical Planning



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# Acknowledgements

- These slides refer to Chapter 3 of the textbook:  
Malik Ghallab, Dana Nau, and Paolo Traverso:  
Automated Planning: Theory and Practice  
Morgan Kaufmann, 2004
- These slides are an adaptation of slides by Dana Nau
- The contributions of these authors are gratefully acknowledged

# 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
- $\text{operator}(\textit{name}, \textit{preconditions}, \textit{effects})$

`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),  $\neg$ empty(crane1),  $\neg$ in(c3,p1),  
           $\neg$ top(c3,p1),  $\neg$ on(c3,c1), top(c1,p1)`

- Classical planning problem:  $P = (\Sigma, s_0, S_g)$

# Review: Set-Theoretic Representation

- Like classical representation, but restricted to propositional logic
- State: a set of propositions - these correspond to ground atoms
  - { on-c-1-pallet, on-c1-r1, on-c1-c2, ..., at-r1-l1, ... }
- 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, on-c3-p1, top-c3-p1, on-c3-p1

*add* : holding-crane1-c3, top-c1-p

- 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:

$load(c, r, l)$

;; robot  $r$  loads container  $c$  at location  $l$

precond:  $rloc(r) = l, cpos(c) = l, rload(r) = nil$

effects:  $rload(r) \leftarrow c, cpos(c) \leftarrow r$

$unload(c, r, l)$

;; 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 practice, state-variable representation is probably more convenient

- Recall that in classical planning even simple problems can have huge search spaces, e.g.
  - States of DWR with 5 locations, 3 piles, 3 robots and 100 containers is  $10^{277}$
  - Largest estimates of particles in universe is only about  $10^{87}$
- How difficult is it to solve classical planning problems?

- 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 analyses are done on 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

- We will consider two language-recognition problems:

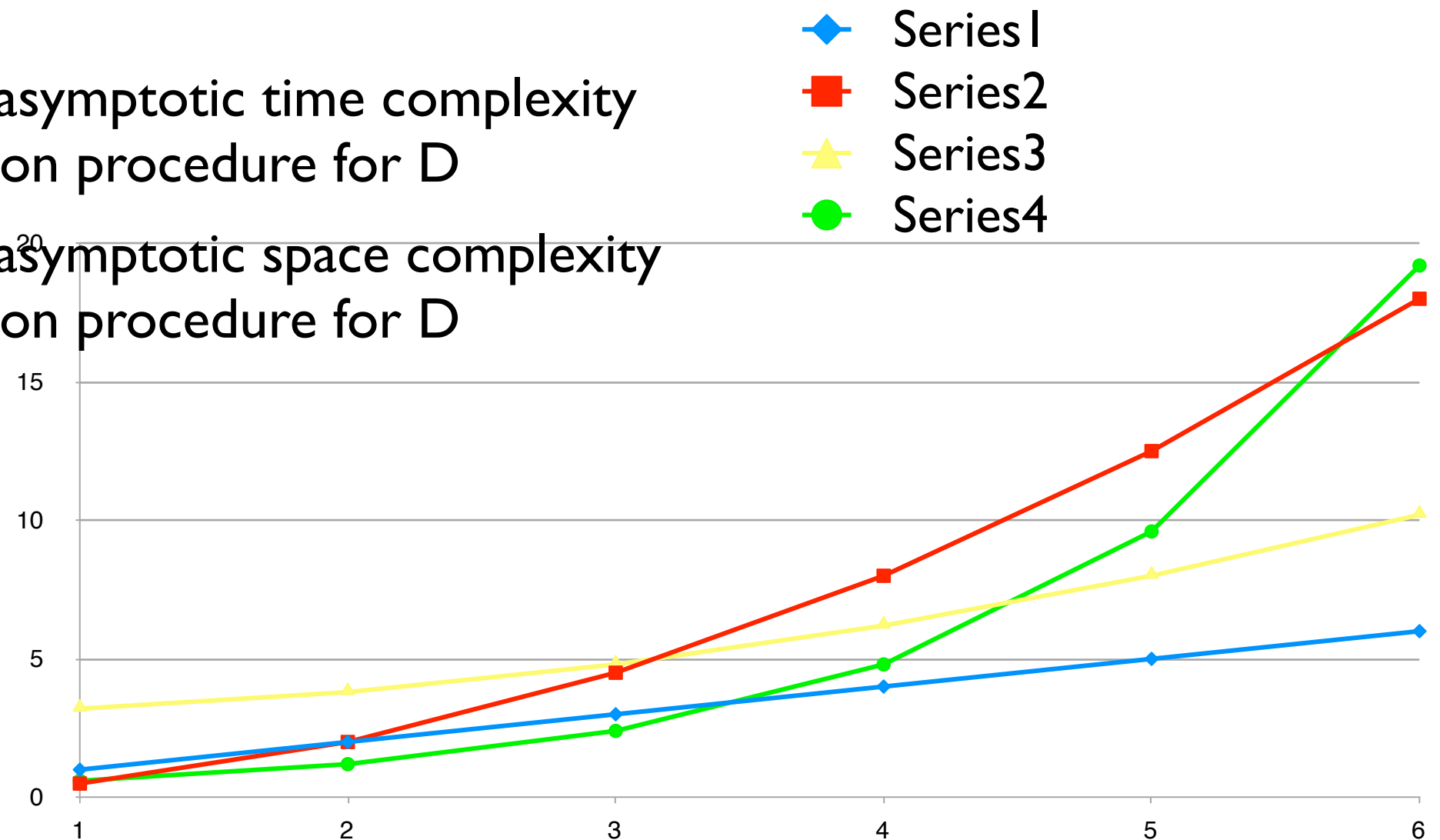
Plan-Existence =  $\{P \mid P \text{ is the statement of a planning problem that has a solution}\}$

Plan-Length =  $\{(P, n) \mid P \text{ is the statement of a planning problem that has a solution of } length \leq n\}$

- Look at complexity of PLAN-EXISTENCE and PLAN-LENGTH under different conditions

# Complexity of Language-Recognition Problems

- Suppose  $R$  is a recognition procedure for  $D$
- Complexity of  $R$ 
  - $T_R(n) =$  worst-case runtime for  $R$  on strings in  $D$  of length  $n$
  - $S_R(n) =$  worst-case space requirement for  $R$  on strings in  $D$  of length  $n$
- Complexity of  $D$ 
  - $T_D =$  best asymptotic time complexity of any recognition procedure for  $D$
  - $S_D =$  best asymptotic space complexity of any recognition procedure for  $D$



# Complexity Classes

$\text{NLOGSPACE} \subseteq$	(nondeterministic procedure, logarithmic space)
$\subseteq P$	(deterministic procedure, polynomial time)
$\subseteq NP$	(nondeterministic procedure, polynomial time)
$\subseteq PSPACE$	(deterministic procedure, polynomial space)
$\subseteq EXPTIME$	(deterministic procedure, exponential time)
$\subseteq NEXPTIME$	(nondeterministic procedure, exponential time)
$\subseteq EXPSPACE$	(deterministic procedure, exponential space)

- Let  $C$  be a complexity class and  $p$  be a language-recognition problem
  - $p$  is  $C$ -hard if for every problem  $q$  in  $C$ ,  
 $q$  can be reduced to  $p$  in a polynomial amount of time
    - NP-hard, PSPACE-hard, etc.
  - $p$  is  $C$ -complete if  $p$  is  $C$ -hard and  $p$  is also in  $C$ 
    - NP-complete, PSPACE-complete, etc.

# 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
- Question marked with \* and answered “yes” take us outside of classical planning

# Decidability of Planning

Halting problem

Can cut off the search at every path  
of length  $n$

Allow function symbols?	Decidability of PLAN-EXISTENCE	Decidability of PLAN-LENGTH
no <sup><math>\alpha</math></sup>	decidable	decidable
yes	semidecidable <sup><math>\beta</math></sup>	decidable

<sup>$\alpha$</sup> This is ordinary classical planning.

<sup>$\beta$</sup> True even if we make several restrictions (see text).

Next: Analyze complexity for the decidable cases

# Complexity of Planning: Classical Representation

PSPACE-complete or  
NP-complete for some sets of  
operators

Kind of representation	How the operators are given	Allow negative effects?	Allow negative preconditions?	Complexity of PLAN-EXISTENCE	Complexity of PLAN-LENGTH
classical rep.	in the input	yes	yes/no	EXPSpace-complete	NEXPTIME-complete
		no	yes	NEXPTIME-complete	NEXPTIME-complete
			no	EXPTIME-complete	NEXPTIME-complete
			$\text{no}^\alpha$	PSPACE-complete	PSPACE-complete
	in advance	yes	yes/no	$\text{PSPACE}^\gamma$	$\text{PSPACE}^\gamma$
		no	yes	$\text{NP}^\gamma$	$\text{NP}^\gamma$
			no	P	$\text{NP}^\gamma$
			$\text{no}^\alpha$	NLOGSPACE	NP

no operator has >1 precondition



# Caveat: Worst-Case Results

- 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 representation	Operations are given	negative effects?	negative preconditions?	PLAN-EXISTENCE	PLAN-LENGTH
classical rep.	in the input	yes	yes/no	EXPSpace-complete	NEXPTIME-complete
		no	yes	NEXPTIME-complete	NEXPTIME-complete
			no	EXPTIME-complete	NEXPTIME-complete
			no <sup>α</sup>	PSPACE-complete	PSPACE-complete
	in advance	yes	yes/no	PSPACE <sup>γ</sup>	PSPACE <sup>γ</sup>
		no	yes	NP <sup>γ</sup>	NP <sup>γ</sup>
			no	P	NP <sup>γ</sup>
			no <sup>α</sup>	NLOGSPACE	NP

# Plan-Length vs Plan-Existence

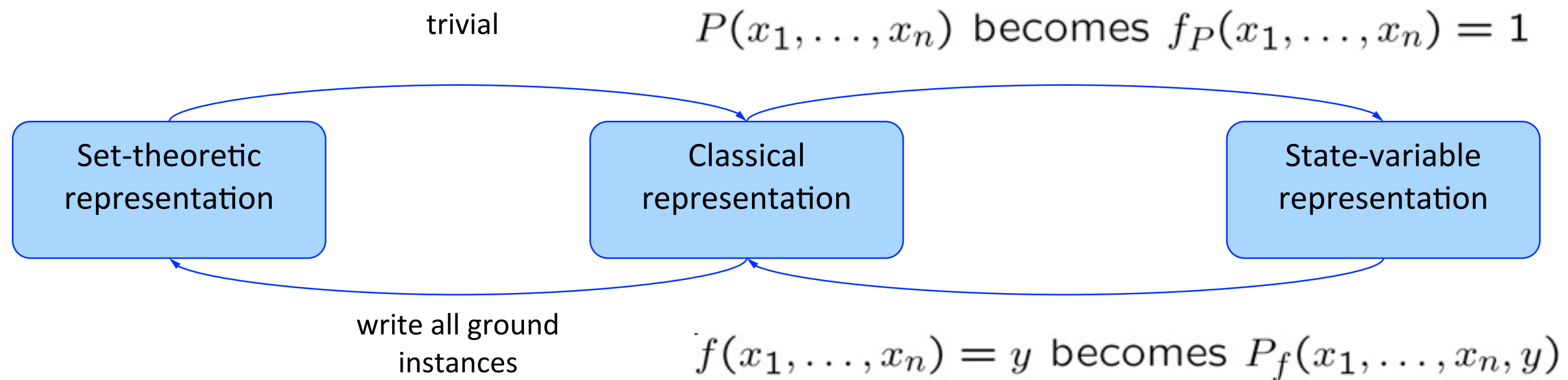
Kind of representation	How the operators are given	Allow negative effects?	Allow negative preconditions?	Complexity of PLAN-EXISTENCE	Complexity of PLAN-LENGTH
classical rep.	in the input	yes	yes/no	EXPSPACE-complete	NEXPTIME-complete
		no	yes	NEXPTIME-complete	NEXPTIME-complete
			no	EXPTIME-complete	NEXPTIME-complete
			no <sup>α</sup>	PSPACE-complete	PSPACE-complete
				PSPACE <sup>γ</sup>	PSPACE <sup>γ</sup>
				NP <sup>γ</sup>	NP <sup>γ</sup>
				P	NP <sup>γ</sup>
			no <sup>α</sup>	NLOGSPACE	NP

- Here, PLAN-LENGTH is easier than PLAN-EXISTENCE for the same reason as in the decidability table
  - Can cut off every search path at depth n



# 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



# Complexity of Planning: Set-Theoretic and State-Variable Representations

no operator has >1 precondition	Kind of representation	How the operators are given	Allow negative effects?	Allow negative preconditions?	Complexity of PLAN-EXISTENCE	Complexity of PLAN-LENGTH
every operator with >1 precondition is the composition of other operators	set-theoretic or ground classical rep.	in the input	yes	yes/no	PSPACE-complete	PSPACE-complete
			no	yes	NP-complete	NP-complete
				no	P	NP-complete
			no <sup>α</sup> /no <sup>β</sup>	no <sup>α</sup> /no <sup>β</sup>	NLOGSPACE-complete	NP-complete
Like classical rep, but fewer lines in the table	state-variable rep.	in the input	yes <sup>δ</sup>	yes/no	EXPSPACE-complete	NEXPTIME-complete
		in advance	yes <sup>δ</sup>	yes/no	PSPACE <sup>γ</sup>	PSPACE <sup>γ</sup>
	ground state-variable rep.	in the input	yes <sup>δ</sup>	yes/no	PSPACE-complete	PSPACE-complete
		in advance	yes <sup>δ</sup>	yes/no	constant time	constant time