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33. Automated Planning: Introduction

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Classification

classification:

Automated Planning

environment:

- **static vs.** dynamic
- ▶ deterministic vs. non-deterministic vs. stochastic
- ► fully vs. partially vs. not observable
- discrete vs. continuous
- ► single-agent vs. multi-agent

problem solving method:

problem-specific vs. general vs. learning

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33.1 Introduction

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Automated Planning

What is Automated Planning?

"Planning is the art and practice of thinking before acting."

— P. Haslum

→ finding plans (sequences of actions) that lead from an initial state to a goal state

our topic in this course: classical planning

- general approach to finding solutions for state-space search problems (Chapters 5–19)
- classical = static, deterministic, fully observable
- variants: probabilistic planning, planning under partial observability, online planning, ...

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Planning: Informally

given:

> state space description in terms of suitable problem description language (planning formalism)

required:

- ▶ a plan, i.e., a solution for the described state space (sequence of actions from initial state to goal)
- or a proof that no plan exists

distinguish between

- optimal planning: guarantee that returned plans are optimal, i.e., have minimal overall cost
- suboptimal planning (satisficing): suboptimal plans are allowed

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Introduction

What is New?

Many previously encountered problems are planning tasks:

- blocks world
- missionaries and cannibals
- ► 15-puzzle

New: we are now interested in general algorithms, i.e., the developer of the search algorithm does not know the tasks that the algorithm needs to solve.

- → no problem-specific heuristics!
- input language to model the planning task

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Automated Planning: Overview

Chapter overview: automated planning

- 33. Introduction
- ▶ 34. Planning Formalisms
- ▶ 35.–36. Planning Heuristics: Delete Relaxation
- ▶ 37. Planning Heuristics: Abstraction
- ▶ 38.–39. Planning Heuristics: Landmarks

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33.2 Repetition: State Spaces

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Repetition: State Spaces

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About This Section

Nothing New Here!

This section is a repetition of Section 5.2 of the chapter "State-Space Search: State Spaces".

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Repetition: State Spaces

Formalization of State Spaces

preliminary remarks:

- to cleanly study search problems we need a formal model
- ► fundamental concept: state spaces
- state spaces are (labeled, directed) graphs
- paths to goal states represent solutions
- shortest paths correspond to optimal solutions

Repetition: State Spaces

State Spaces

Definition (state space)

A state space or transition system is a 6-tuple

 $S = \langle S, A, cost, T, s_0, S_{\star} \rangle$ with

- S: finite set of states
- A: finite set of actions
- ightharpoonup cost: $A o \mathbb{R}_0^+$ action costs
- $T \subseteq S \times A \times S$ transition relation; deterministic in $\langle s, a \rangle$ (see next slide)
- ▶ $s_0 \in S$ initial state
- $ightharpoonup S_{\star} \subseteq S$ set of goal states

German: Zustandsraum, Transitionssystem, Zustände, Aktionen, Aktionskosten, Transitions-/Übergangsrelation, deterministisch, Anfangszustand, Zielzustände

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State Spaces: Transitions, Determinism

Definition (transition, deterministic)

Let $S = \langle S, A, cost, T, s_0, S_{\star} \rangle$ be a state space.

The triples $\langle s, a, s' \rangle \in T$ are called (state) transitions.

We say S has the transition $\langle s, a, s' \rangle$ if $\langle s, a, s' \rangle \in T$.

We write this as $s \xrightarrow{a} s'$, or $s \rightarrow s'$ when a does not matter.

Transitions are deterministic in $\langle s, a \rangle$: it is forbidden to have both $s \stackrel{a}{\to} s_1$ and $s \stackrel{a}{\to} s_2$ with $s_1 \neq s_2$.

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Compact Descriptions

33.3 Compact Descriptions

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Repetition: State Spaces

State Spaces: Terminology

terminology:

- predecessor, successor
- applicable action
- path, length, costs
- reachable
- solution, optimal solution

German: Vorgänger, Nachfolger, anwendbare Aktion, Pfad, Länge, Kosten, erreichbar, Lösung, optimale Lösung

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State Spaces with Declarative Representations

How do we represent state spaces in the computer?

previously: as black box

now: as declarative description

reminder: Chapter 6

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State Spaces with Declarative Representations

represent state spaces declaratively:

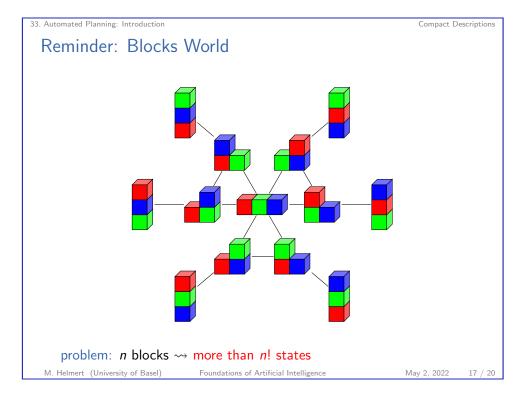
- **compact** description of state space as input to algorithms
- algorithms directly operate on compact description
- → allows automatic reasoning about problem: reformulation, simplification, abstraction, etc.

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33.4 Summary

33. Automated Planning: Introduction

Compact Descriptions

Compact Description of State Spaces

How to describe state spaces compactly?

Compact Description of Several States

- ▶ introduce state variables
- > states: assignments to state variables
- \rightarrow e.g., *n* binary state variables can describe 2^n states
- ► transitions and goal are compactly described with a logic-based formalism

different variants: different planning formalisms

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Summar

Summary

- planning: search in general state spaces
- input: compact, declarative description of state space

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