

Foundations of Artificial Intelligence

33. Automated Planning: Introduction

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33.1 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

33.1 Introduction

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, ...

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

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

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**

33.2 Repetition: State Spaces

About This Section

Nothing New Here!

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

State Spaces

To cleanly study search problems we need a **formal model**.

Definition (state space)

A **state space** or **transition system** is a 6-tuple $\mathcal{S} = \langle S, A, cost, T, s_I, S_\star \rangle$ with

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

State Spaces: Terminology & Notation

Definition (transition, deterministic)

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

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

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

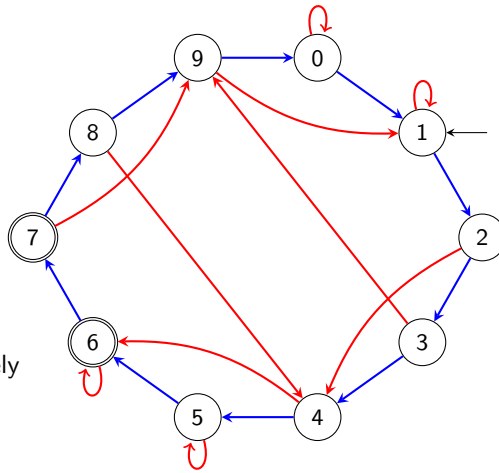
We write this as $s \xrightarrow{a} s'$, or $s \rightarrow s'$ when a does not matter.

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

Graph Interpretation

state spaces are often depicted as **directed, labeled graphs**

- ▶ **states**: graph vertices
- ▶ **transitions**: labeled arcs
(here: colors instead of labels)
- ▶ **initial state**: incoming arrow
- ▶ **goal states**: double circles
- ▶ **actions**: the arc labels
- ▶ **action costs**: described separately
(or implicitly = 1)



State Spaces: Terminology

terminology:

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

33.3 Compact Descriptions

State Spaces with Declarative Representations

How do we represent state spaces in the computer?

previously: as black box

now: as **declarative description**

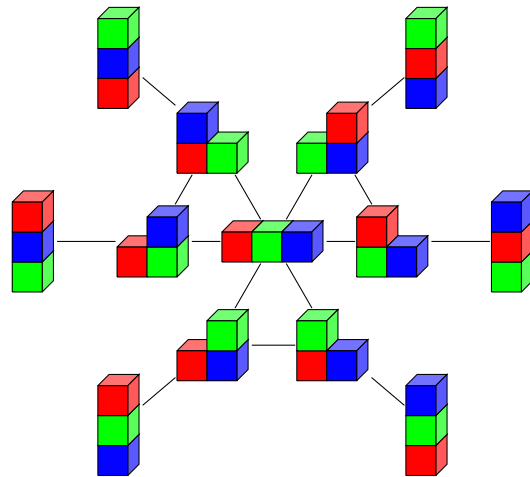
reminder: Chapter 6

State Spaces with Declarative Representations

represent state spaces **declaratively**:

- ▶ **compact** description of state space as input to algorithms
 \rightsquigarrow state spaces **exponentially larger** than the input
- ▶ algorithms directly operate on compact description
 \rightsquigarrow allows automatic reasoning about problem:
 reformulation, simplification, abstraction, etc.

Reminder: Blocks World



problem: n blocks \rightsquigarrow more than $n!$ states

Compact Description of State Spaces

How to describe state spaces compactly?

Compact Description of Several States

- ▶ introduce **state variables**
- ▶ states: assignments to state variables
- ~ 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**

33.4 Summary

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

Summary