

Basics of AI and Machine Learning

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

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

Here: **classical planning**

- **general** approach to finding solutions
for **state-space search problems**
- **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
- route planning in romania
- 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

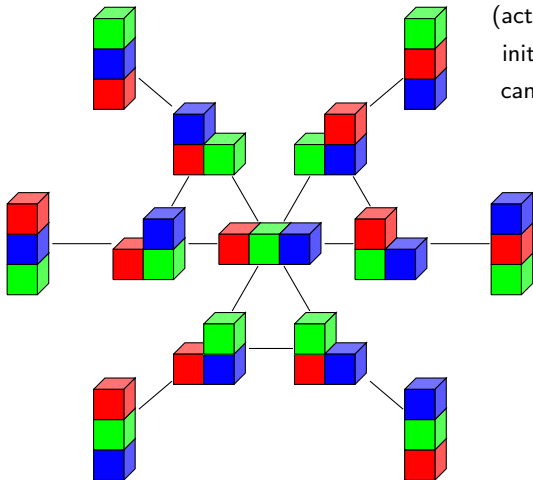
Automated Planning: Overview

Chapter overview: automated planning

- **Introduction**
- The STRIPS Planning Formalism
- Other Planning Formalisms
- Planning Heuristics
- Alternatives to Heuristic Search

Compact Descriptions

Reminder: State Space



(action names omitted;
initial state and goal
can be arbitrary)

- state spaces are (labeled, directed) graphs
- **terminology:** predecessor, successor, applicable action, path, length, costs, reachable, solution, optimal solution

State Spaces with Declarative Representations

How do we represent state spaces in the computer?

previously: as black box

now: as **declarative description**

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.

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**

Summary

Summary

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