# Foundations of Artificial Intelligence <br> B2. State-Space Search: Representation of State Spaces 

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## State-Space Search: Overview

Chapter overview: state-space search

- B1-B3. Foundations
- B1. State Spaces
- B2. Representation of State Spaces
- B3. Examples of State Spaces
- B4-B8. Basic Algorithms
- B9-B15. Heuristic Algorithms

Representation of State Spaces

## Representation of State Spaces

- practically interesting state spaces are often huge $\left(10^{10}, 10^{20}, 10^{100}\right.$ states)
- How do we represent them, so that we can efficiently deal with them algorithmically?
three main options:
(1) as explicit (directed) graphs
(2) with declarative representations
(3) as a black box

German: explizit, deklarativ, Black Box

## Explicit Graphs

## State Spaces as Explicit Graphs

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represent state spaces as explicit directed graphs:

- vertices $=$ states
- directed arcs $=$ transitions
$\rightsquigarrow$ represented as adjacency list or adjacency matrix
German: Adjazenzliste, Adjazenzmatrix

Example (explicit graph for bounded inc-and-square)
ai-b02-bounded-inc-and-square.graph

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Example (explicit graph for 8-puzzle)
ai-b02-puzzle8.graph

## State Spaces as Explicit Graphs: Discussion

## discussion:

- impossible for large state spaces (too much space required)
- if spaces small enough for explicit representations, solutions easy to compute: Dijkstra's algorithm $O(|S| \log |S|+|T|)$
- interesting for time-critical all-pairs-shortest-path queries (examples: route planning, path planning in video games)


## Declarative Representations

## State Spaces with Declarative Representations

## 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.

Example (declarative representation for 8-puzzle)
puzzle8-domain.pddl + puzzle8-problem.pddl

## Black Box

## State Spaces as Black Boxes

## State Spaces as Black Boxes

Define an abstract interface for state spaces.
For state space $\mathcal{S}=\left\langle S, A, \operatorname{cost}, T, s_{\mathrm{I}}, S_{\mathrm{G}}\right\rangle$
we need these methods:

- init(): generate initial state result: state $s$
- is_goal(s): test if $s$ is a goal state result: true if $s \in S_{\mathrm{G}}$; false otherwise
- $\operatorname{succ}(s)$ : generate applicable actions and successors of $s$ result: sequence of pairs $\left\langle a, s^{\prime}\right\rangle$ with $s \xrightarrow{a} s^{\prime}$
- cost(a): gives cost of action a result: $\operatorname{cost}(a)\left(\in \mathbb{N}_{0}\right)$

Remark: we will extend the interface later in a small but important way

## State Spaces as Black Boxes: Example and Discussion

## Example (Black Box Representation for 8-Puzzle) <br> demo: puzzle8.py

- in the following: focus on black box model
- explicit graphs only as illustrating examples
- near end of semester: declarative state spaces (classical planning)


## Summary

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- state spaces often huge ( $>10^{10}$ states) $\rightsquigarrow$ how to represent?
- explicit graphs: adjacency lists or matrices; only suitable for small problems
- declaratively: compact description as input to search algorithms
- black box: implement an abstract interface

