

# Foundations of Artificial Intelligence

## B9. State-Space Search: Heuristics

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March 16, 2026

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## State-Space Search: Overview

### Chapter overview: state-space search

- ▶ B1–B3. Foundations
- ▶ B4–B8. Basic Algorithms
- ▶ B9–B15. Heuristic Algorithms
  - ▶ **B9. Heuristics**
  - ▶ B10. Analysis of Heuristics
  - ▶ B11. Best-first Graph Search
  - ▶ B12. Greedy Best-first Search, A\*, Weighted A\*
  - ▶ B13. IDA\*
  - ▶ B14. Properties of A\*, Part I
  - ▶ B15. Properties of A\*, Part II

## B9.1 Introduction

## Informed Search Algorithms

search algorithms considered so far:

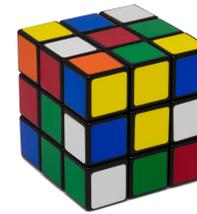
example:  $b = 13$ ;  $10^5$  nodes/second

- ▶ **uninformed** (“blind”): use **no information** besides **formal definition** to solve a problem
- ▶ **scale poorly**: prohibitive time (and space) requirements for seemingly **simple** problems (time complexity usually  $O(b^d)$ )

$d$	nodes	time
4	30 940	0.3 s
6	$5.2 \cdot 10^6$	52 s
8	$8.8 \cdot 10^8$	147 min
10	$10^{11}$	17 days
12	$10^{13}$	8 years
14	$10^{15}$	1 352 years
16	$10^{17}$	$2.2 \cdot 10^5$ years
18	$10^{20}$	$38 \cdot 10^6$ years

## Informed Search Algorithms

Rubik's cube:



- ▶ branching factor:  $\approx 13$
- ▶ typical solution length: 18

search algorithms considered now:

- ▶ **idea**: try to find (problem-specific) criteria to distinguish **good** and **bad states**
- ▶ **heuristic** (“informed”) search algorithms prefer **good states**

Richard Korf, Finding Optimal Solutions to Rubik's Cube Using Pattern Databases (AAAI, 1997)

## B9.2 Heuristics

### Heuristics

#### Definition (heuristic)

Let  $S$  be a state space with states  $S$ .

A **heuristic function** or **heuristic** for  $S$  is a function

$$h : S \rightarrow \mathbb{R}_0^+ \cup \{\infty\},$$

mapping each state to a nonnegative number (or  $\infty$ ).

## Heuristics: Intuition

**idea:**  $h(s)$  estimates distance (= cost of cheapest path) from  $s$  to closest goal state

- ▶ heuristics can be **arbitrary** functions
- ▶ **intuition:**
  - ① the closer  $h$  is to true goal distance, the more efficient the search using  $h$
  - ② the better  $h$  separates states that are **close** to the goal from states that are **far**, the more efficient the search using  $h$

## Why “Heuristic”?

**What does “heuristic” mean?**

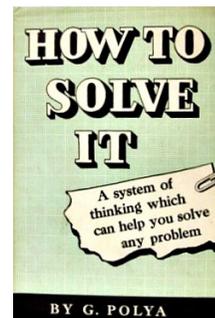
- ▶ from ancient Greek  $\acute{\epsilon}\nu\text{ρ}\text{ι}\sigma\text{κ}\omega$  (= I find)
- ▶ same origin as  $\acute{\epsilon}\nu\text{ρ}\eta\text{κ}\alpha!$



## Why “Heuristic”?

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- ▶ popularized by George Pólya: How to Solve It (1945)
- ▶ in computer science often used for: rule of thumb, inexact algorithm
- ▶ in state-space search technical term for **goal distance estimator**



## Representation of Heuristics

In our black box model, heuristics are an additional element of the state space interface:

### State Spaces as Black Boxes (Extended)

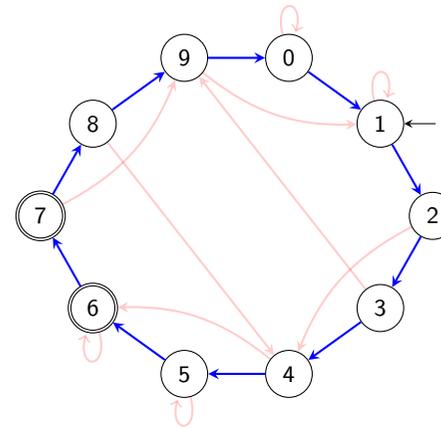
- ▶ `init()`
- ▶ `is_goal(s)`
- ▶ `succ(s)`
- ▶ `cost(a)`
- ▶ `h(s)`: heuristic value for state  $s$   
result: nonnegative integer or  $\infty$

# B9.3 Examples

## Bounded Inc-and-Square

bounded inc-and-square:

possible heuristics:



$$h_1(s) = \begin{cases} 0 & \text{if } s = 7 \\ (16 - s) \bmod 10 & \text{otherwise} \end{cases}$$

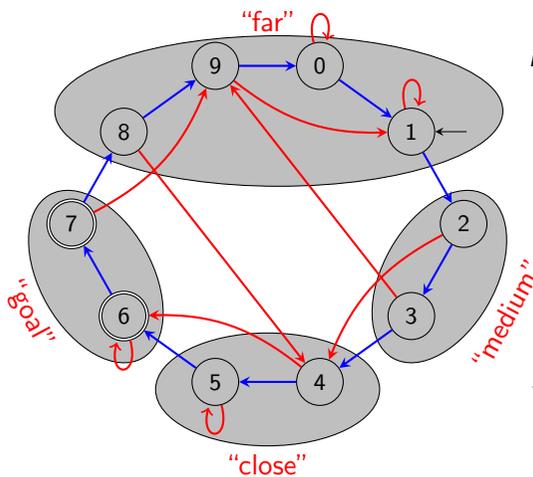
↪ number of *inc* actions to goal

How accurate is this heuristic?

## Bounded Inc-and-Square

bounded inc-and-square:

possible heuristics:



$$h_1(s) = \begin{cases} 0 & \text{if } s = 7 \\ (16 - s) \bmod 10 & \text{otherwise} \end{cases}$$

↪ number of *inc* actions to goal

$$h_2(s) = \begin{cases} 0 & \text{if } s \text{ is a "goal"} \\ 1 & \text{if } s \text{ is "close"} \\ 2 & \text{if } s \text{ is "medium"} \\ 3 & \text{if } s \text{ is "far"} \end{cases}$$

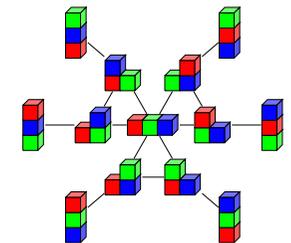
↪ categorize states

How accurate is this heuristic?

## Example: Blocks World

possible heuristic:

count blocks  $x$  that currently lie on  $y$  and must lie on  $z \neq y$  in the goal (including case where  $y$  or  $z$  is the table)

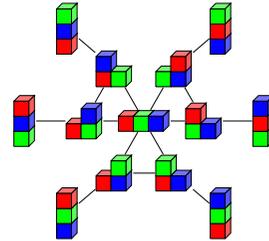


## Example: Blocks World

possible heuristic:

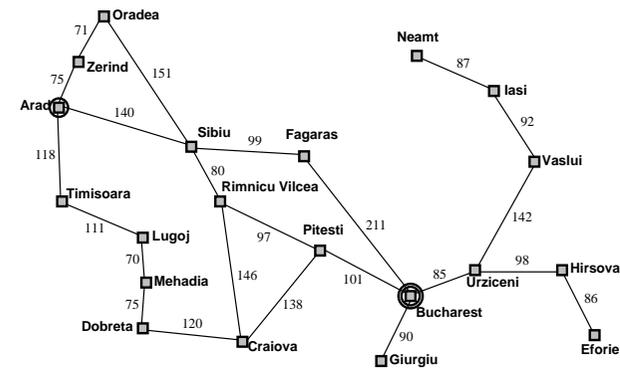
count blocks  $x$  that currently lie on  $y$  and must lie on  $z \neq y$  in the goal (including case where  $y$  or  $z$  is the table)

How accurate is this heuristic?



## Example: Route Planning in Romania

possible heuristic: straight-line distance to Bucharest



## Example: Missionaries and Cannibals

Setting: Missionaries and Cannibals

- ▶ Six people must cross a river.
- ▶ Their rowing boat can carry one or two people across the river at a time (it is too small for three).
- ▶ Three people are missionaries, three are cannibals.
- ▶ Missionaries may never stay with a majority of cannibals.

possible heuristic: number of people on the wrong river bank

↪ with our formulation of states as triples  $\langle m, c, b \rangle$ :

$$h(\langle m, c, b \rangle) = m + c$$

## B9.4 Summary

## Summary

- ▶ **heuristics** estimate distance of a state to the goal
- ▶ can be used to **focus** search on **promising** states
- ↔ **soon**: search algorithms that use heuristics