

# Foundations of Artificial Intelligence

## B9. State-Space Search: Heuristics

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B9.1 Introduction

B9.2 Heuristics

B9.3 Examples

B9.4 Summary

# 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:



search algorithms considered now:

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

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

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

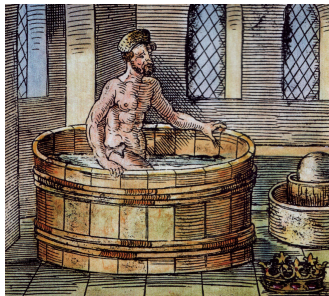
**idea:**  $h(s)$  estimates distance (= cost of cheapest path) from  $s$  to closest goal state

- ▶ 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?

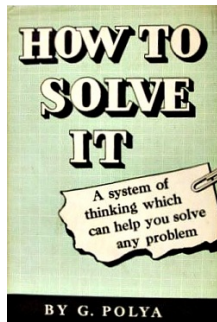
- ▶ from ancient Greek εύρισκω (= I find)
- ▶ same origin as εύρηκα!



# Why “Heuristic”?

## What does “heuristic” mean?

- ▶ from ancient Greek  $\epsilon\upsilon\text{ρισκ}\omega$  (= I find)
- ▶ same origin as  $\epsilon\upsilon\text{ρηκα!}$
- ▶ 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:

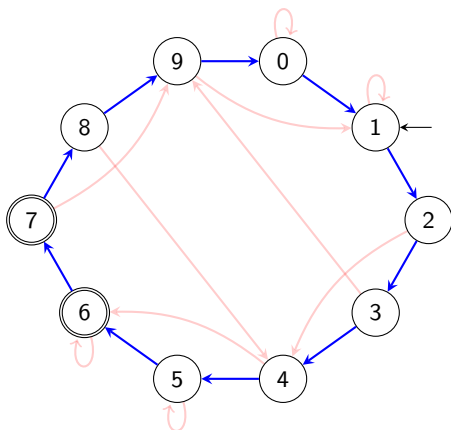
## State Spaces as Black Boxes (Extended)

- ▶ `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:

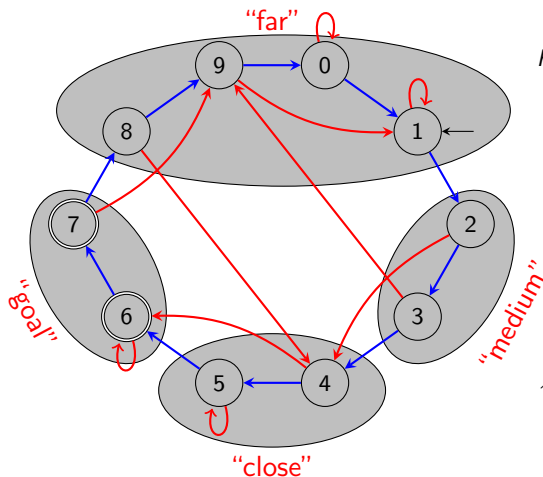
$$h_1(s) = \begin{cases} 0 & \text{if } s = 7 \\ (16 - s) \bmod 10 & \text{otherwise} \end{cases}$$

$\rightsquigarrow$  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

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

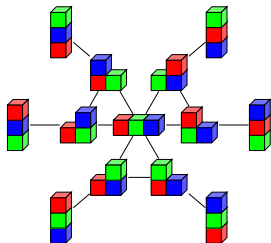
↪ **categorize** states

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)



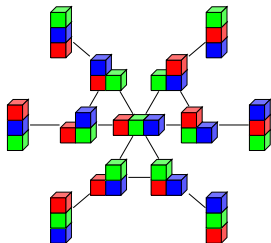


# Example: Blocks World

possible heuristic:

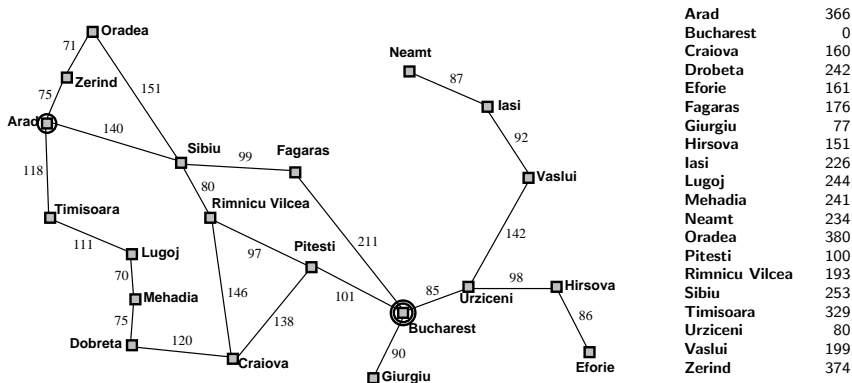
count blocks  $x$  that currently lie on  $y$   
and must lie on  $z \neq y$  in the goal  
(including case where  $y$  or  $z$  is the table)

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