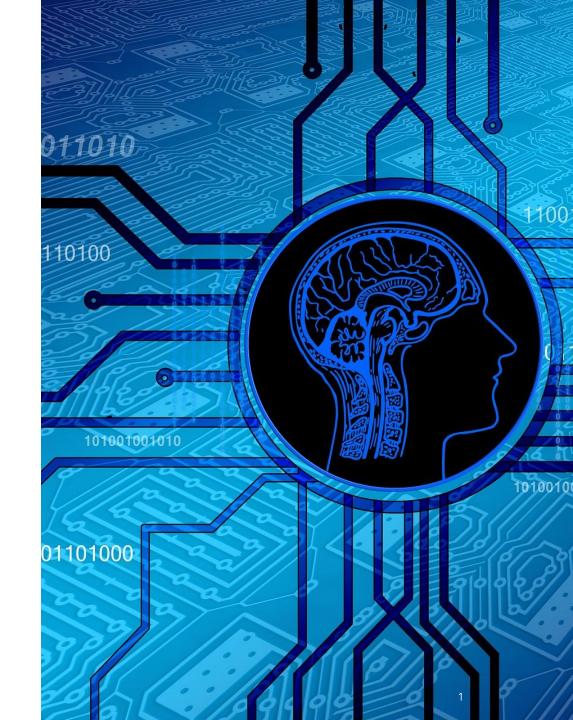
Informed Search

Informatics 2D: Reasoning and Agents Lecture 4

Adapted from slides provided by Dr Petros Papapanagiotou



Criterion	Breadth- First	Uniform- Cost	Depth- First	Depth- Limited	Iterative Deepening	Bidirectional (if applicable)
Complete? Time Space Optimal?	$\begin{array}{c} \operatorname{Yes}^a \\ O(b^d) \\ O(b^d) \\ \operatorname{Yes}^c \end{array}$	Yes ^{a,b} $O(b^{1+\lfloor C^*/\epsilon \rfloor})$ $O(b^{1+\lfloor C^*/\epsilon \rfloor})$ Yes	$egin{array}{c} {\sf No} \\ O(b^m) \\ O(bm) \\ {\sf No} \end{array}$	$egin{array}{c} \operatorname{No} & \ O(b^\ell) & \ O(b\ell) & \ \operatorname{No} & \ \end{array}$	Yes ^a $O(b^d)$ O(bd) Yes ^c	Yes a,d $O(b^{d/2})$ $O(b^{d/2})$ Yes c,d
_						

Review: Summary of algorithms

Review: Tree search

function TREE-SEARCH(problem) returns a solution, or failure

initialize the frontier using the initial state of *problem*

loop do

if the frontier is empty then return failure

choose a leaf node and remove it from the frontier

if the node contains a goal state **then return** the corresponding solution

expand the chosen node, adding the resulting nodes to the frontier

A search strategy is defined by picking the order of node expansion from the **frontier.**

Review: Graph search

function GRAPH-SEARCH(problem) returns a solution, or failure
initialize the frontier using the initial state of problem
initialize the explored set to be empty
loop do
if the frontier is empty then return failure
choose a leaf node and remove it from the frontier
if the node contains a goal state then return the corresponding solution
add the node to the explored set
expand the chosen node, adding the resulting nodes to the frontier
only if not in the frontier or explored set

A search strategy is defined by picking the order of node expansion from the **frontier.**

Making search 'informed'

- Tree-Search
- Graph Search

What if we **order** the nodes in the frontier by decreasing *desirability*?

A search strategy is defined by picking the order of node expansion from the **frontier.**

Best-first search

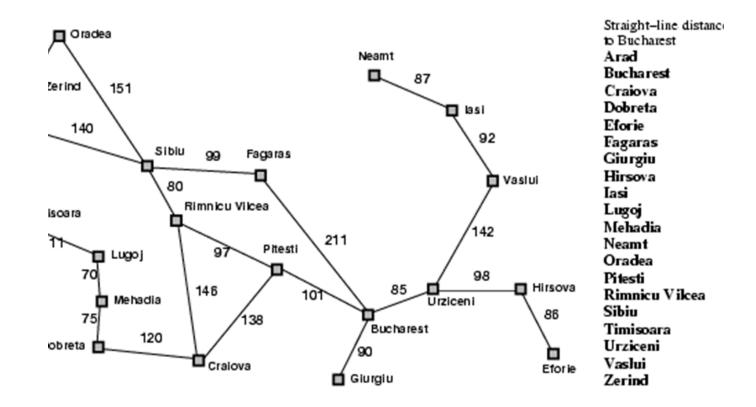
An instance of general TREE-SEARCH or GRAPH-SEARCH

→ Use an evaluation function f(n) for each node n
 • estimate of "desirability"

→ Expand most desirable unexpanded node, usually the node with the lowest evaluation

Heuristics

- Any method that is believed or practically proven to be useful for the solution of a given problem.
 - No guarantee that it will always work or lead to an optimal solution!
- We use heuristics to guide search.
 - This may not change the worst-case complexity of the algorithm, but <u>can help in the average case</u>.
- We will introduce admissibility, consistency conditions to identify good heuristics.



Romania Example



• Evaluation function f(n) = h(n) (heuristic)

• estimated cost of cheapest path from state at node *n* to a *goal* state

• e.g., $h_{SLD}(n)$ = straight-line distance from *n* to Bucharest

• Greedy best-first search expands the node that appears to be closest to goal.

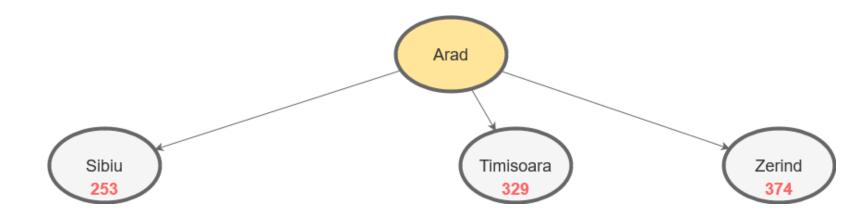
Arad				



Straight-line distan	ce
to Bucharest	
Arad	366
Bucharest	0
Craiova	160
Dobreta	242
Eforie	161
Fagaras	176
Giurgiu	77
Hirsova	151
Iasi	226
Lugoj	244
Mehadia	241
Neamt	234
Oradea	380
Pitesti	10
Rimnicu Vilcea	193
Sibiu	253
Timisoara	329
Urziceni	30
Vaslui	199
Zerind	374

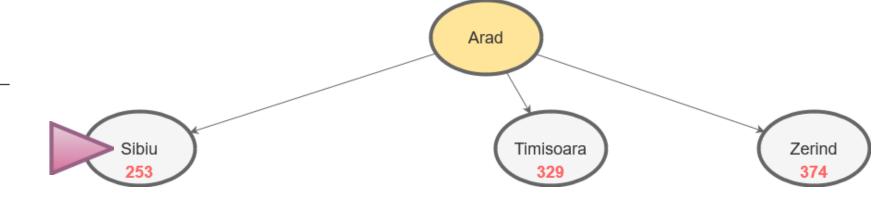
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Straight-line distan	ce
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Oradea	390
Pitesti	10
Rimnicu Vikea	193
Sibiu	253
Timisoara	329
Urziceni	30
Vaslui	199
Zerind	374



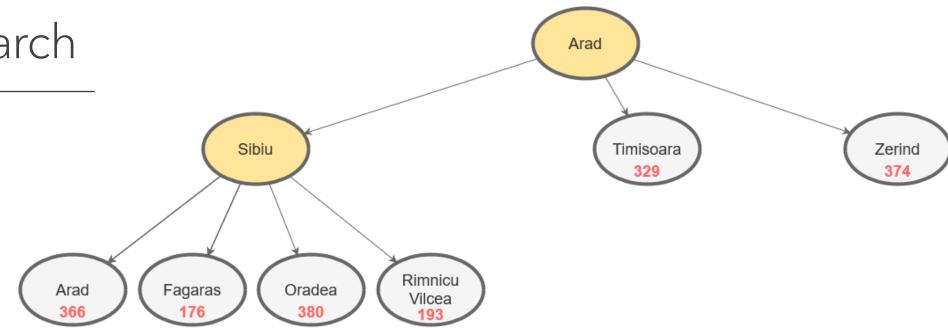
Sibiu	Timiso- ara	Zerind					
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Straight-line distance to Bucharest Arad 366 Bucharest 0 Craiova 160 Dobreta 242 Eforie 161 Fagaras 176 Giurgiu 77 Hirsova 151 Iasi 226 Lugoj 244 Mehadia 241 Neamt 234 Oradea 390 Pitesti 10 Rimnicu Vikea 193 Sibiu 253 Timisoara 329 Urziceni 30 Vaslui 199 Zerind 374



Timiso- ara	Zerind	Arad	Fagaras	Oradea	RV			
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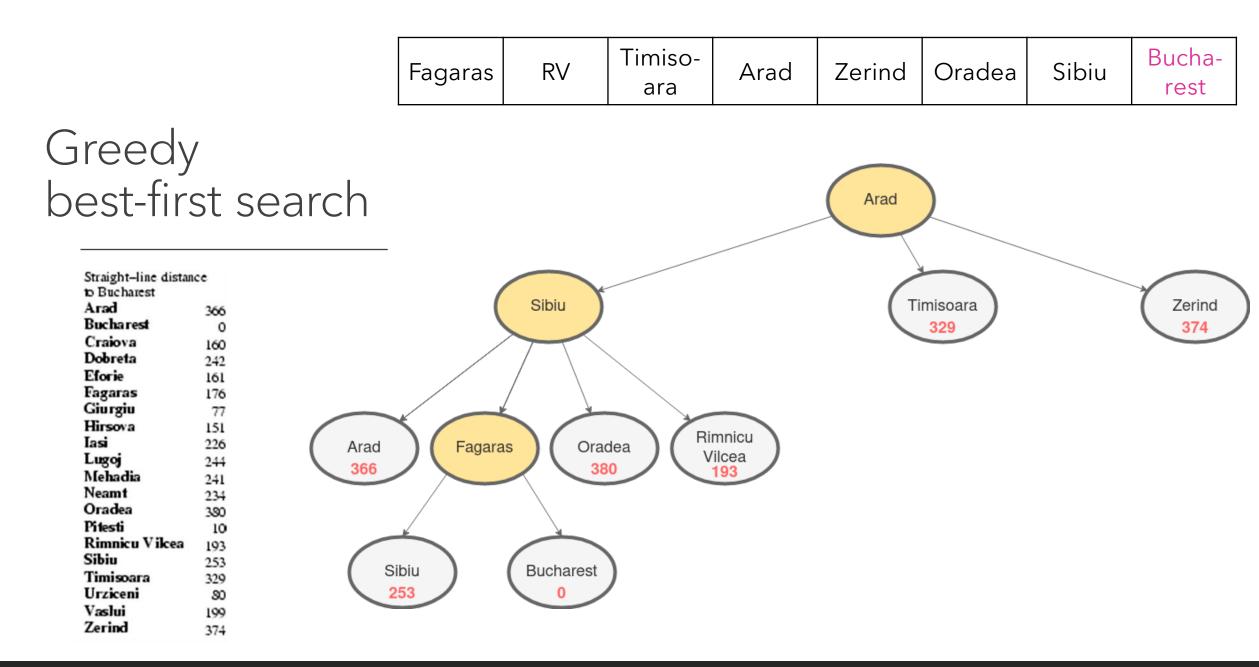
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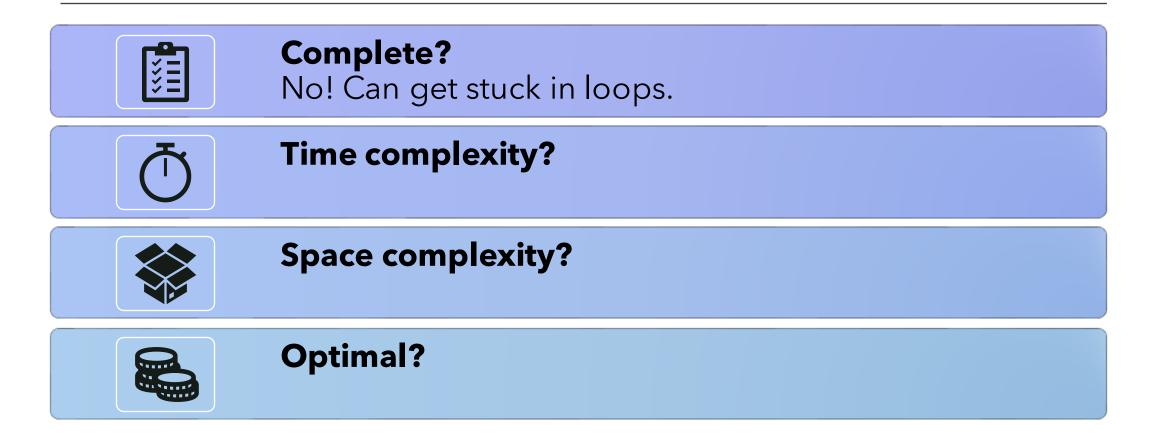
			Fagaras	RV	Timiso- ara	Arad	Zerind	Oradea	
Greedy best-firs	, st sea	arch					Arad		
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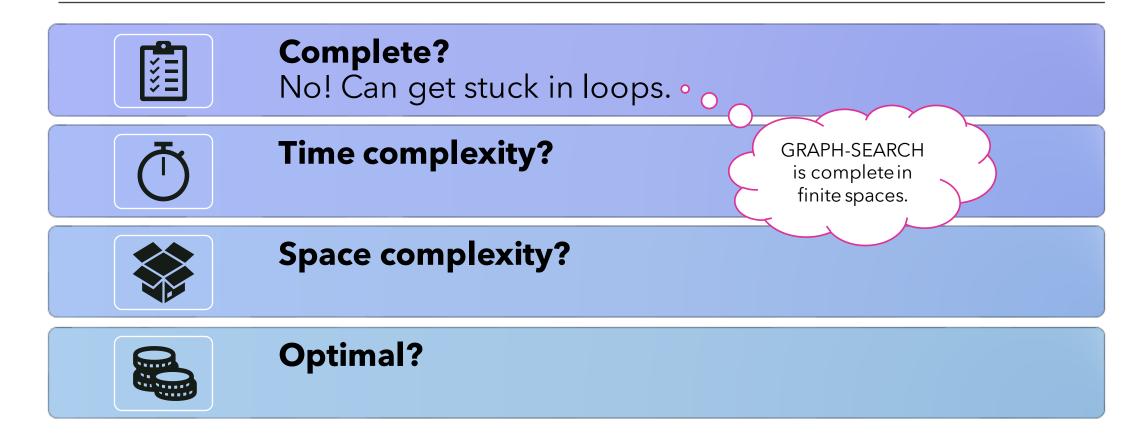
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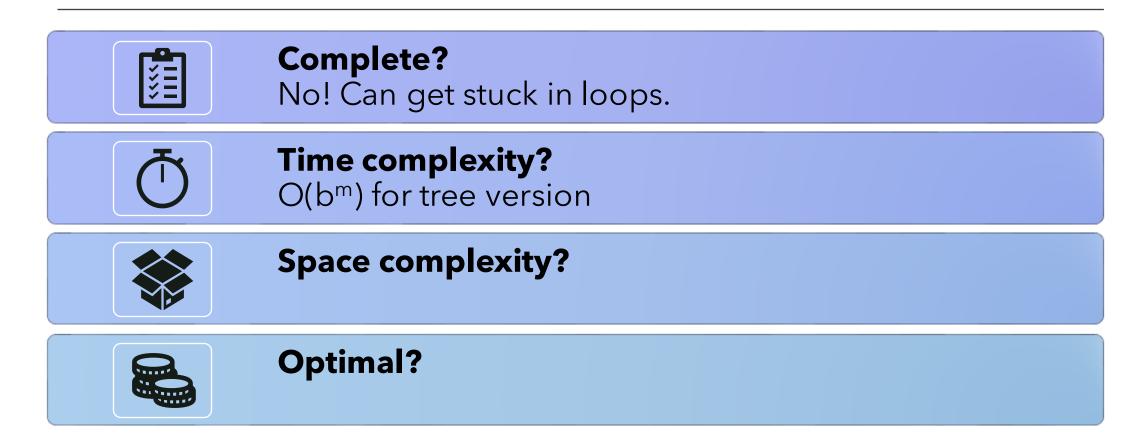
INF2D: REASONING AND AGENTS

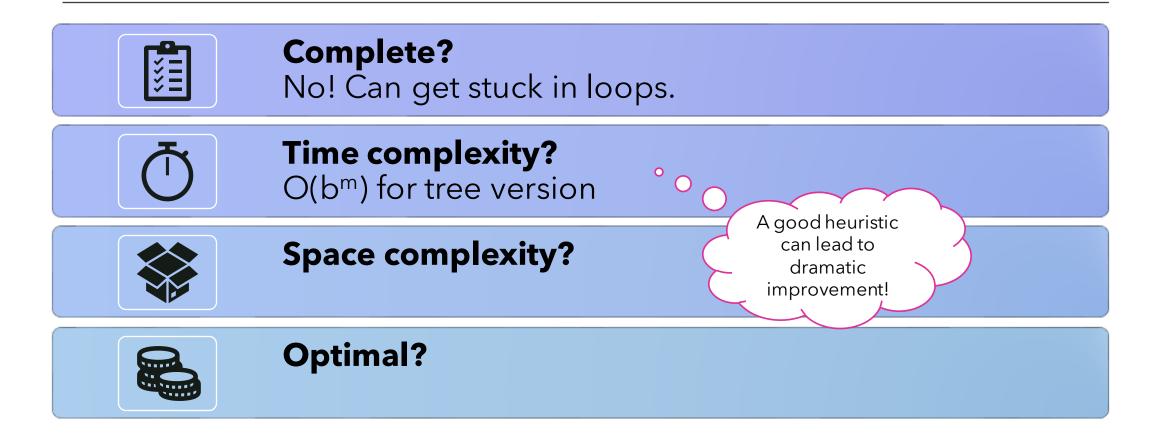


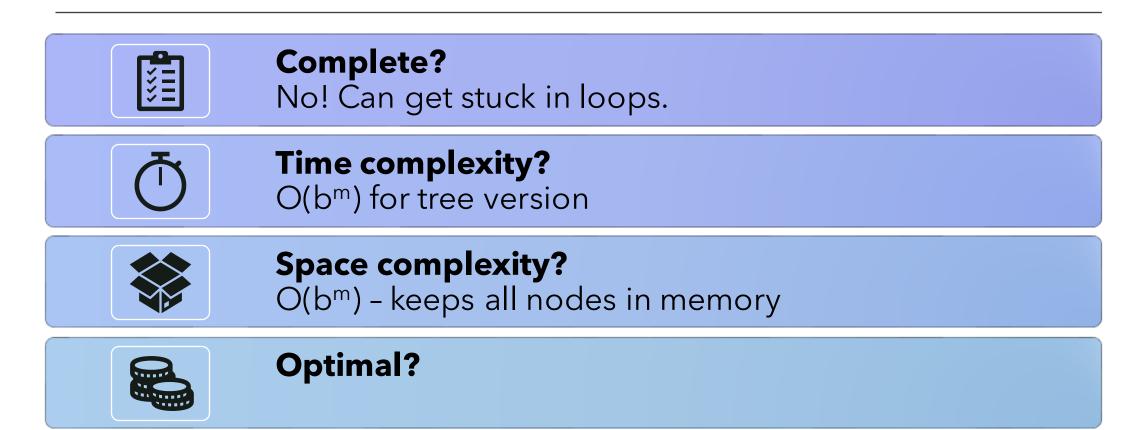


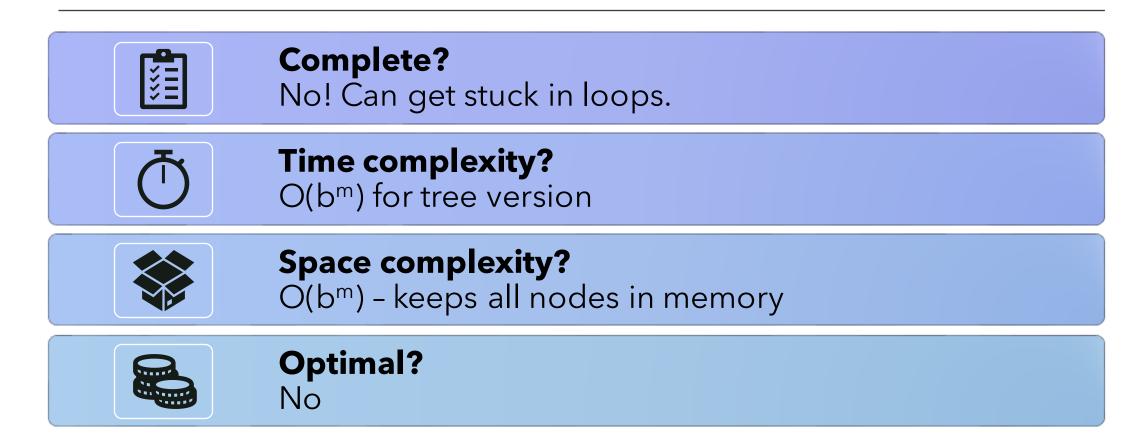










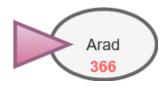


A* Search

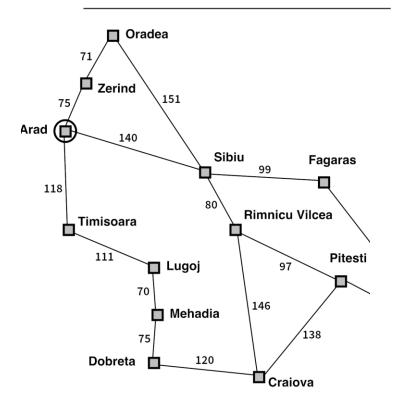
A* search

- Evaluation function f(n) = g(n) + h(n)
 - g(n) = cost so far to reach n
 - h(n) = estimated cost from n to goal
 - f(n) = estimated total cost of path through n to goal

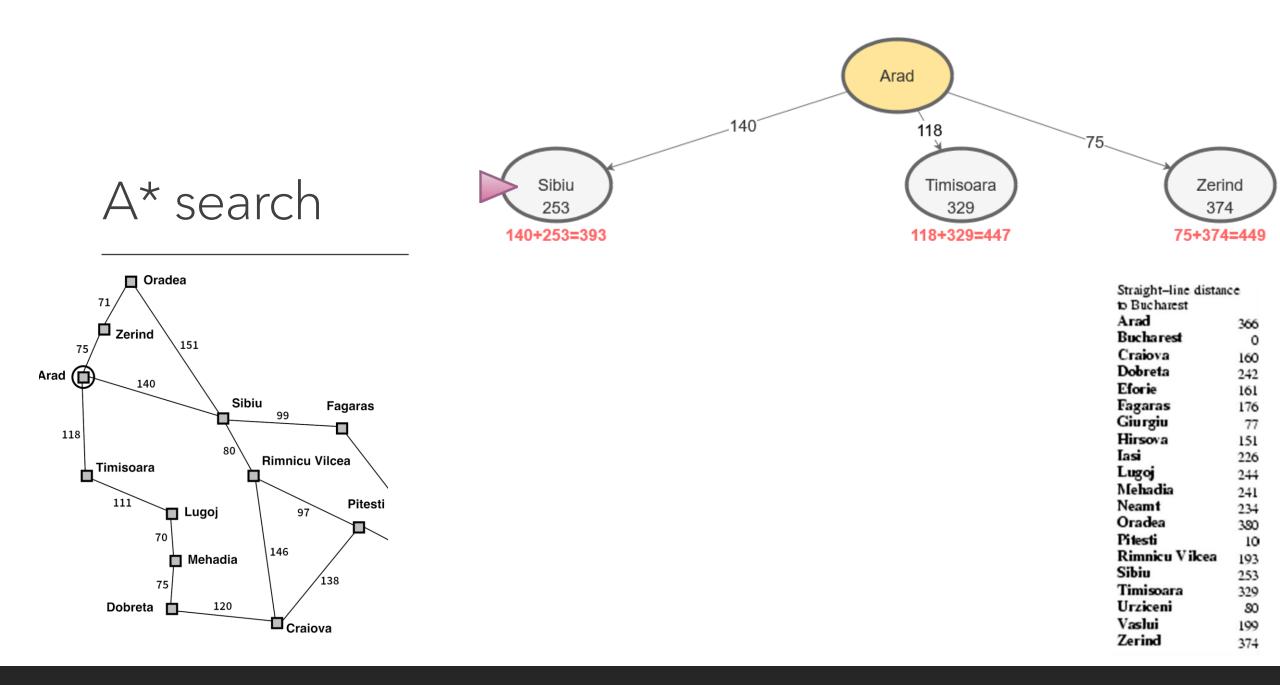
• Avoid expanding paths that are already expensive

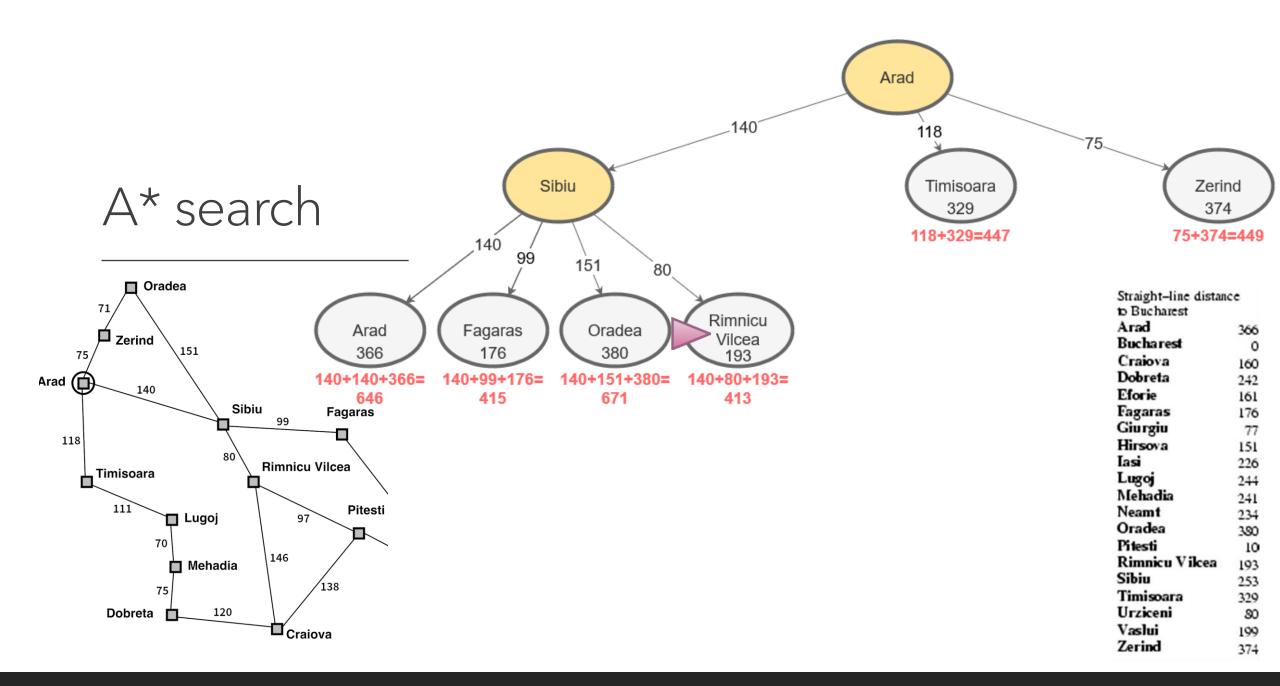


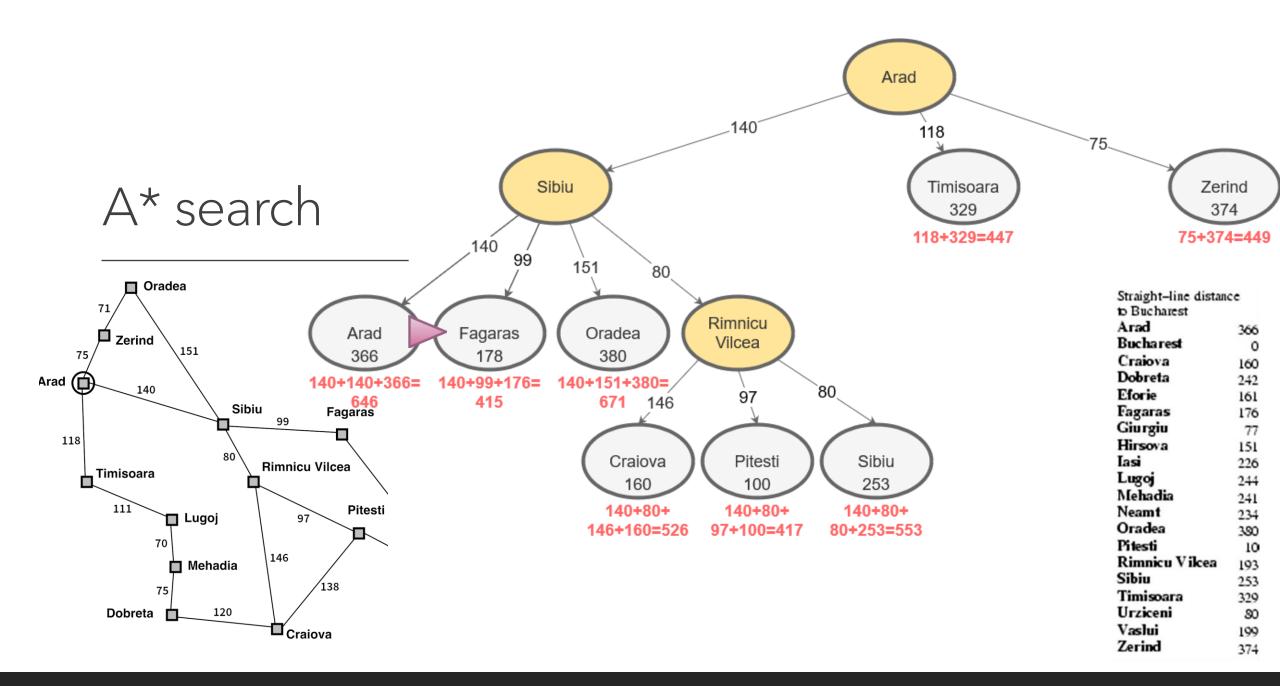
A* search

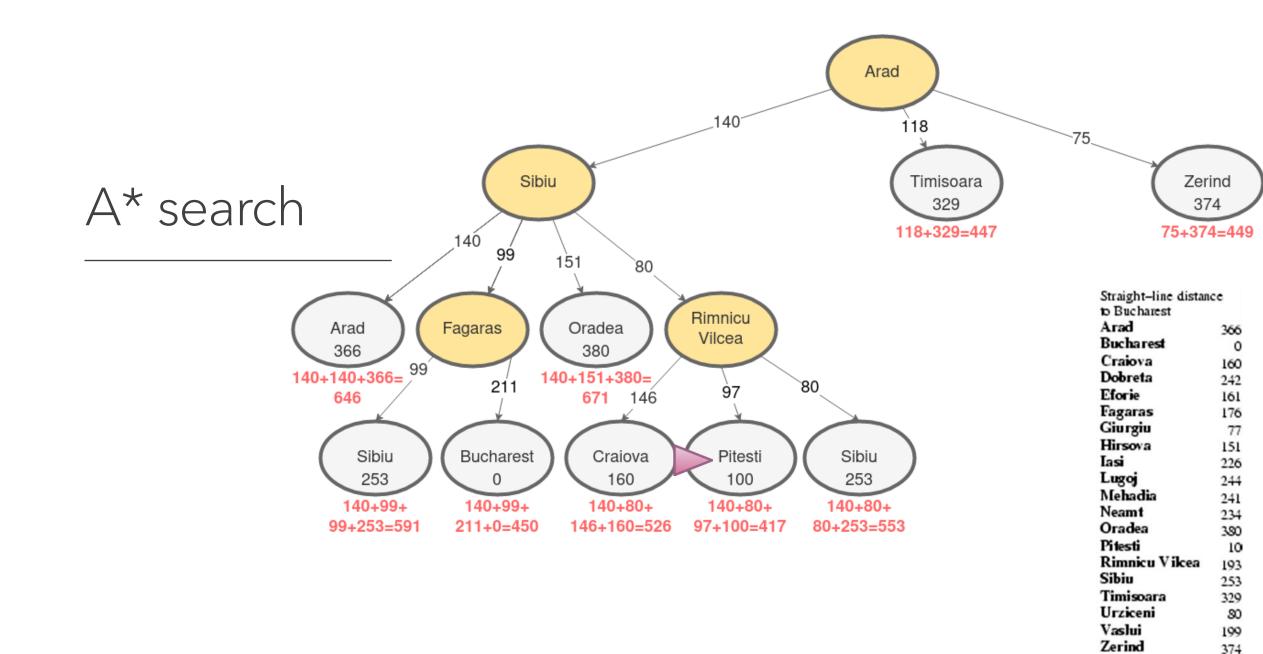


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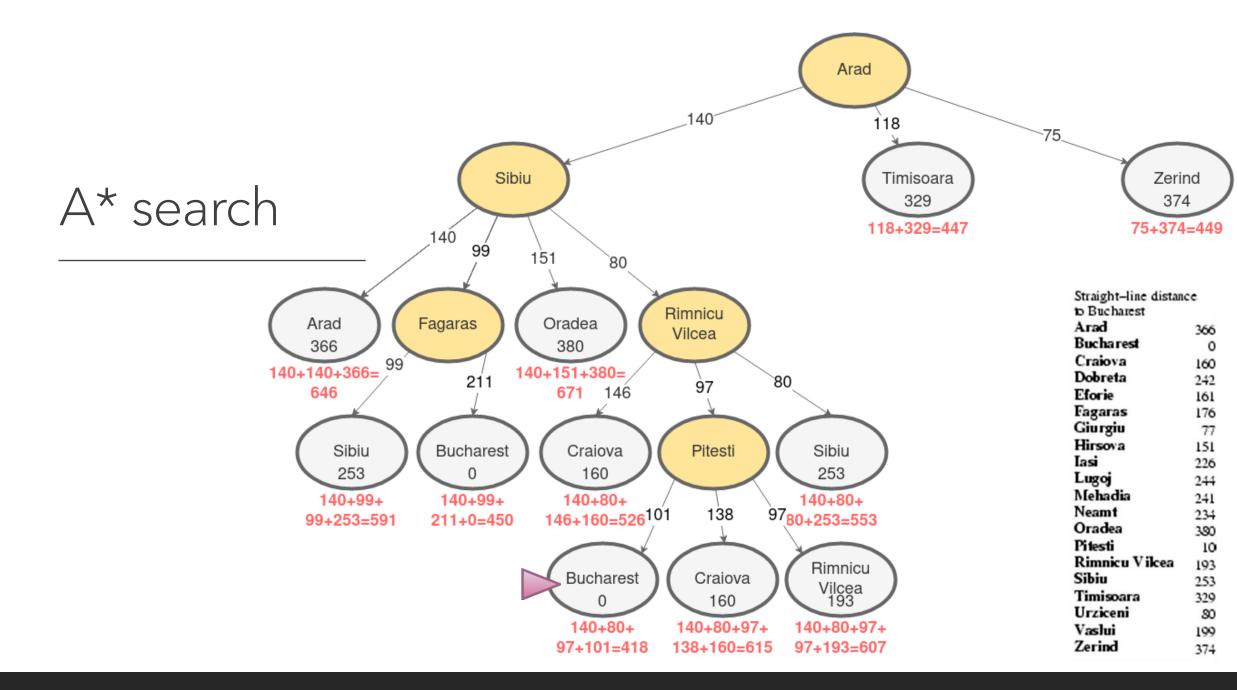








INF2D: REASONING AND AGENTS



Heuristics

Admissible heuristics

• A heuristic *h(n)* is admissible if for every node *n*:

h(n) ≤ **h*(n)**

where **h***(**n**) is the true cost to reach the goal state from *n*.

• An admissible heuristic never overestimates the cost to reach the goal, i.e., it is **optimistic**

Example: $h_{SLD}(n)$ (never overestimates the actual road distance)

Admissible heuristic = optimal A*

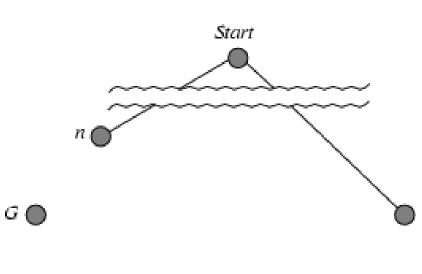
h(n) never overestimates the cost to reach the goal

Thus, f(n) = g(n) + h(n) never overestimates the true cost of a solution

THEOREM

If h(n) is admissible, A^{*} using **TREE-SEARCH** is optimal

Proof: Optimality of A*

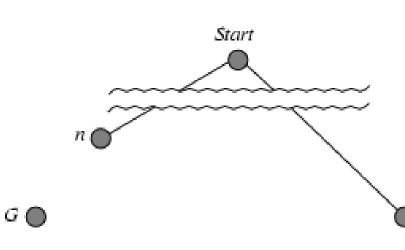


Suppose some suboptimal goal G_2 has been generated and is in the frontier.

Let *n* be an unexpanded node in the frontier such that *n* is on a shortest path to an optimal goal *G*.

 $f(G_2) = g(G_2) \qquad \text{since } h(G_2) = 0$ $G_2 \quad g(G_2) > g(G) \qquad \text{since } G_2 \text{ is suboptimal}$ $f(G) = g(G) \qquad \text{since } h(G) = 0$ $f(G_2) > f(G) \qquad \text{from above}$

Proof: Optimality of A*



Suppose some suboptimal goal G_2 has been generated and is in the frontier.

Let *n* be an unexpanded node in the frontier such that *n* is on a shortest path to an optimal goal *G*.

 $f(G) < f(G_2)$ from above

 G_2 $h(n) \leq h^*(n)$ since h is admissible

 $g(n)+h(n)\leq g(n)+h^*(n)$

 $f(n) \leq f(G)$

Hence $f(n) < f(G_2)$, and A* will never select G2 for expansion

Consistent heuristics

A heuristic h(n) is consistent if for every node n, every successor n' of n generated by any action a,

 $h(n) \le c(n, a, n') + h(n')$

If *h* is consistent, we have

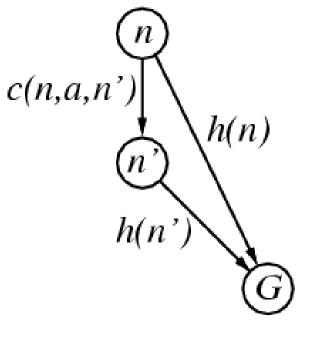
$$f(n') = g(n') + h(n')$$

= $g(n) + c(n, a, n') + h(n')$
 $\ge g(n) + h(n)$
 $\ge f(n)$

i.e., f(n) is non-decreasing along any path.

THEOREM

If h(n) is consistent, A^{*} using **GRAPH-SEARCH** is optimal

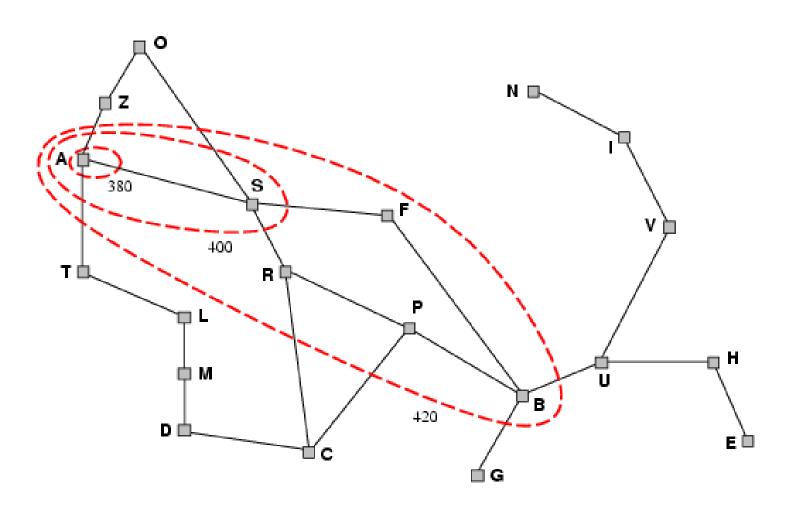


Optimality of A*

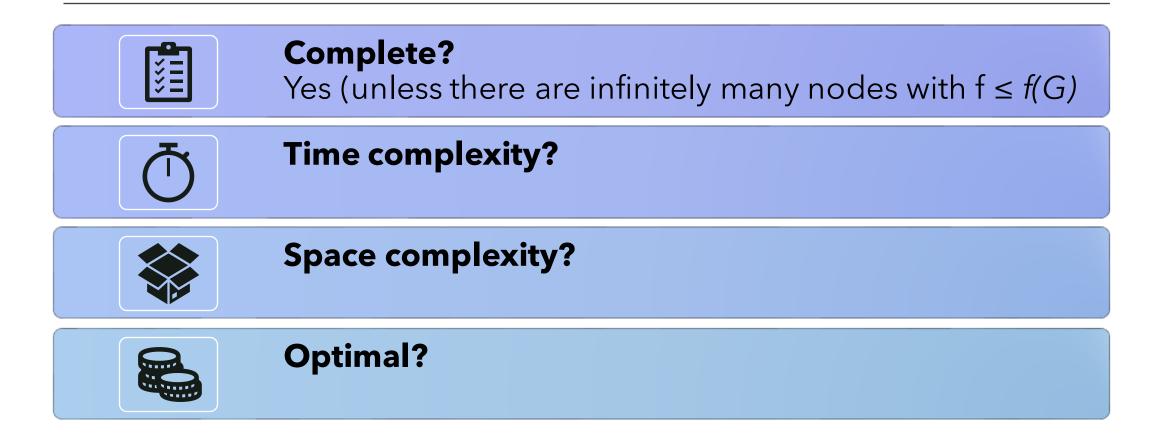
A^{*} expands nodes in order of increasing *f* value

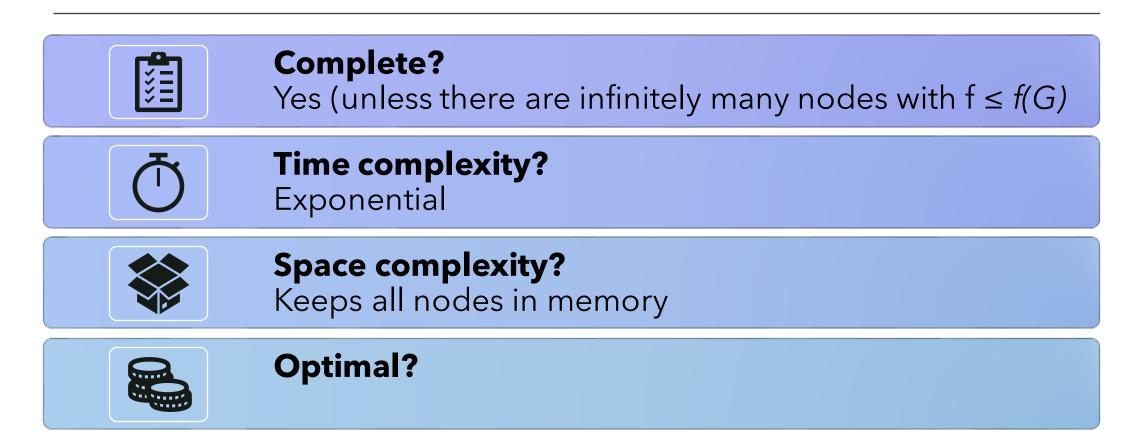
Gradually adds "*f*-contours" of nodes

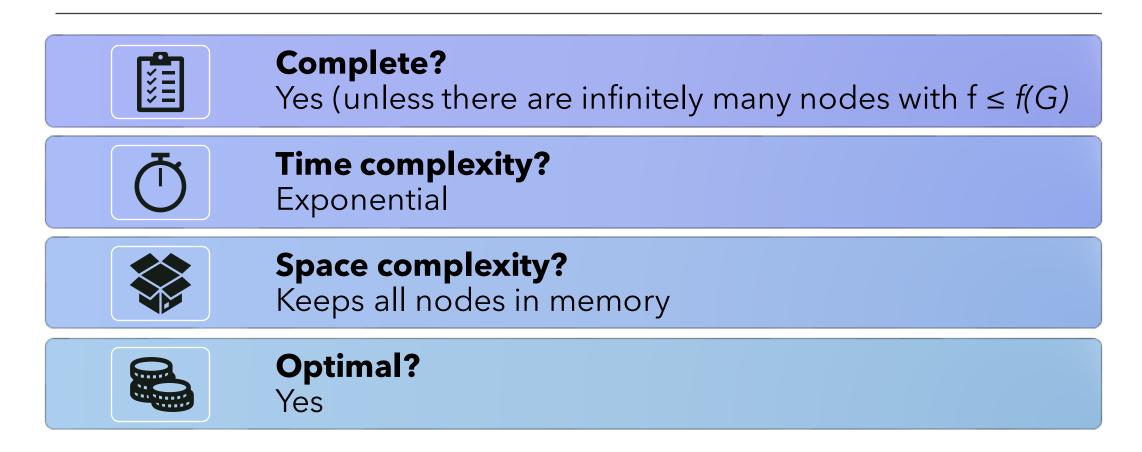
Contour *i* has all nodes with $f=f_i$, where $f_i < f_{i+1}$



	Complete?
Ō	Time complexity?
	Space complexity?
	Optimal?







Admissible heuristics

Example: 8-puzzle:

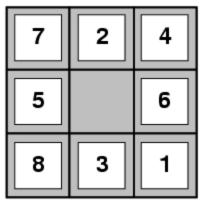
 $\circ h_1(n) =$ number of misplaced tiles

• $h_2(n)$ = total Manhattan distance

(i.e., no. of squares from desired location of each tile)

$$h_1(S) = ?$$

$$h_2(S) = ?$$





4		
6	3	
1	6	



Goal State

2

5

8

Relaxed problems

- A problem with <u>fewer restrictions</u> on the actions is called a *relaxed problem*.
- The cost of an optimal solution to a relaxed problem is an *admissible heuristic* for the original problem.
- If the rules of the 8-puzzle are relaxed so that a tile can move <u>anywhere</u>,
 then h₁(n) gives the shortest solution
- If the rules are relaxed so that a tile can move to <u>any adjacent square</u>,
 then h₂(n) gives the shortest solution

Use relaxation to automatically generate admissible heuristics!



Dominance

- If $h_2(n) \ge h_1(n)$ for all n (both admissible) then
 - h₂ dominates h1
 - \circ h₂ is better for search
- Typical search costs (average number of nodes expanded):
 - d=12 IDS = 3,644,035 nodes A*(h₁) = 227 nodes A*(h₂) = 73 nodes
 - \circ d=24 IDS ≈ 54,000,000,000 nodes A*(h₁) = 39,135 nodes A*(h₂) = 1,641 nodes

Summary

Smart search based on heuristic scores.

- Best-first search
- Greedy best-first search
- \circ A^{*} search
- Admissible heuristics and optimality.

Why?

- Informed search allows us to use domain knowledge to our advantage.
- Optimality over some utility can often be the top priority.
- A* is **very** popular! (e.g., pathfinding)
- A* is simple, yet very efficient.
- A* is too good sometimes (e.g., in games).