

Artificial Intelligence

For HEDSPI Project

Lecturer 4 - Search

Lecturers :

Le Thanh Huong

Tran Duc Khanh

Dept of Information Systems

Faculty of Information Technology - HUT

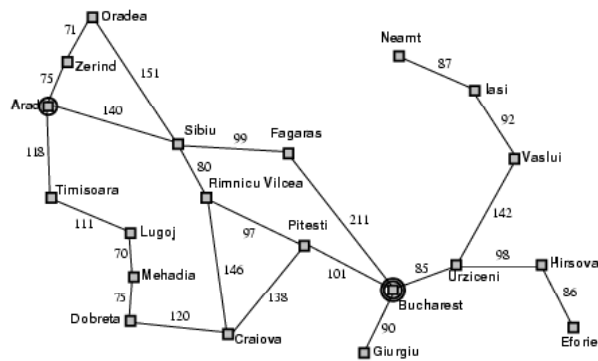
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Outline

- Graph search
- Best-first search
- A* search

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Graph search

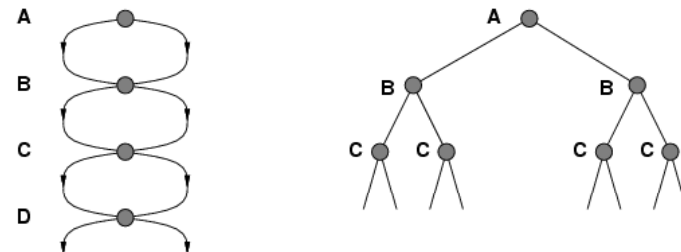


Get from Arad to Bucharest as quickly as possible

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Graph search

- Failure to detect repeated states can turn a linear problem into an exponential one!



- Very simple fix: never expand a node twice

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Graph search

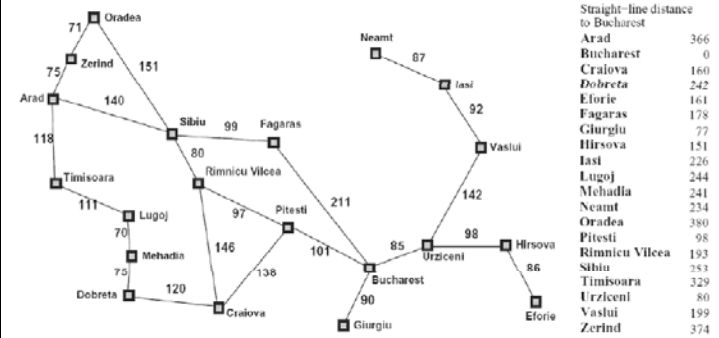
```

function Graph-Search(problem, fringe) returns a solution, or failure
  fringe ← Insert(Make-Node(Initial-State(problem)), fringe);
  closed ← an empty set
  while (fringe not empty)
    node ← RemoveFirst(fringe);
    if (Goal-Test(problem, State(node))) then return Solution(node);
    if (State(node) is not in closed then
      add State(node) to closed
      fringe ← InsertAll(Expand(node, problem), fringe);
    end if
  end
  return failure;
  
```

- Never expand a node twice!

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Straight Line Distances

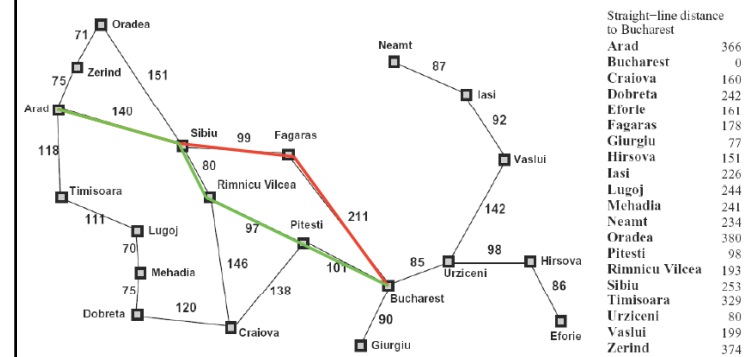


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Best-first search

- Idea: use an **evaluation function** $f(n)$ for each node
 - estimate of "desirability"
 - Expand most desirable unexpanded node
- Order the nodes in fringe in decreasing order of desirability
- Special cases:
 - greedy best-first search
 - A* search

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Greedy Best-First Search

- Evaluation function $f(n) = h(n)$ (heuristic)
= estimate of cost from n to *goal*
- 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

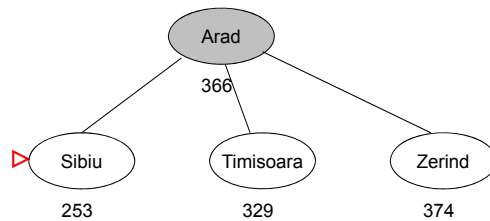
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Greedy best-first search example



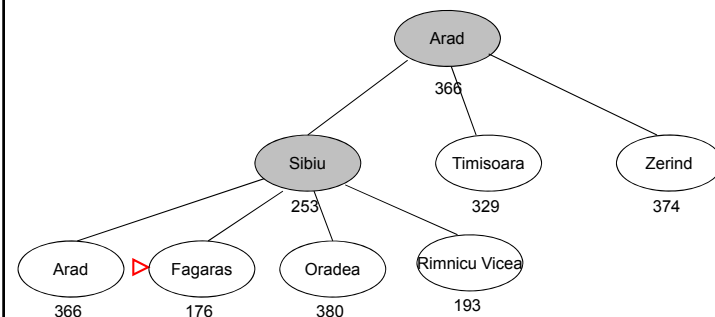
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Greedy best-first search example



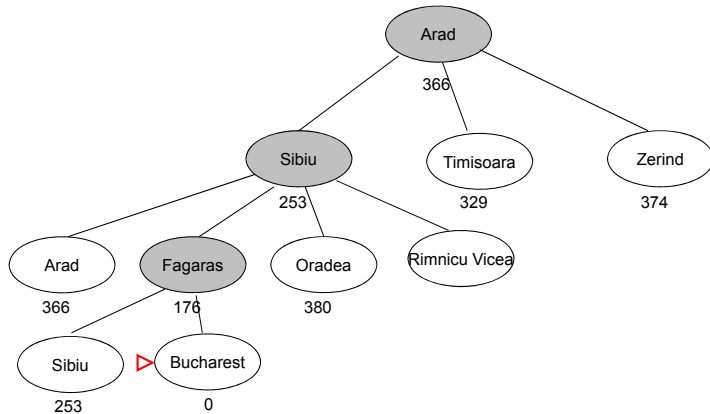
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Greedy best-first search example



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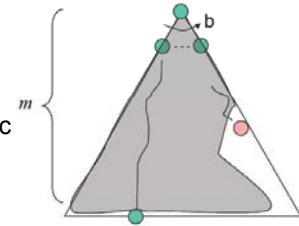
Greedy best-first search example



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Greedy Best-First Search

- **Complete?** No – can get stuck in loops, e.g., Iasi → Neamt → Iasi → Neamt → ...
 - **Time?** $O(b^m)$, but a good heuristic can give dramatic improvement
 - **Space?** $O(b^m)$ -- keeps all nodes in memory
 - **Optimal?** No
- ⇒ A* search
- Can we make it optimal? → No



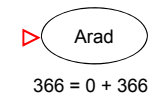
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A* search

- Idea: Expand unexpanded node with lowest evaluation value
- 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
- Nodes are ordered according to $f(n)$.

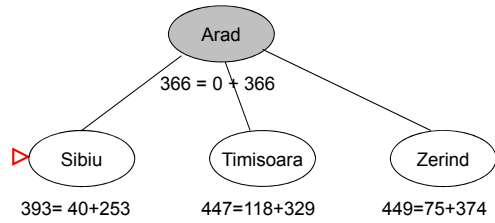
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A* search example



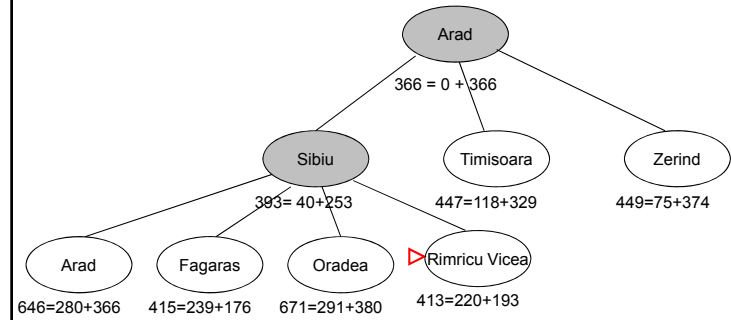
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A* search example



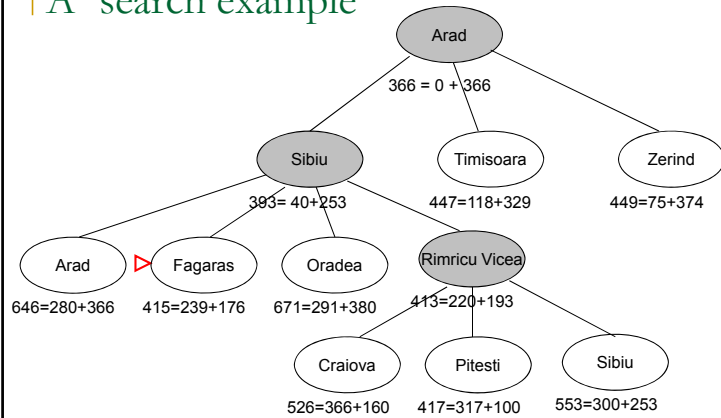
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A* search example



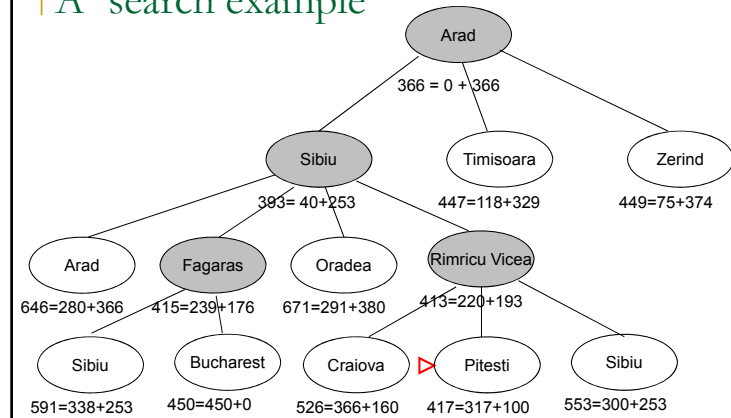
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A* search example

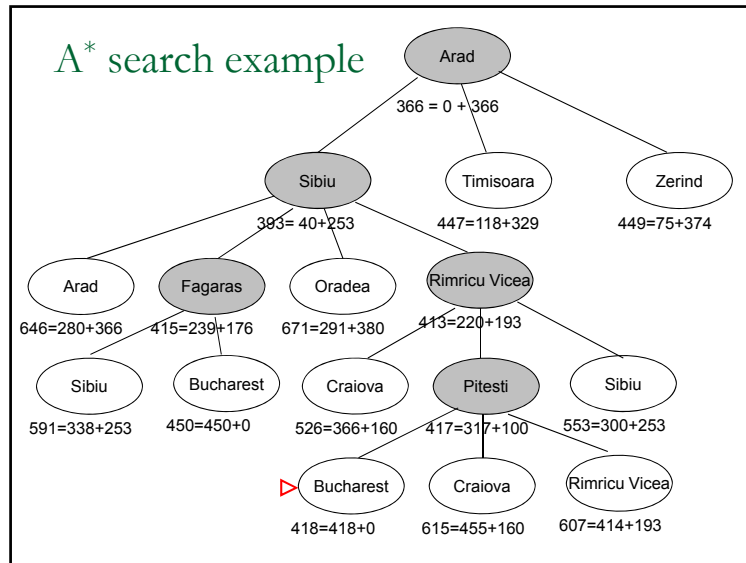


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A* search example



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Can we Prove Anything?

- If the state space is finite and we avoid repeated states, the search is complete, but in general is not optimal
- If the state space is finite and we do not avoid repeated states, the search is in general not complete
- If the state space is infinite, the search is in general not complete

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Admissible heuristic

- Let $h^*(N)$ be the **true** cost of the optimal path from N to a goal node
- Heuristic $h(N)$ is **admissible** if:

$$0 \leq h(N) \leq h^*(N)$$
- An admissible heuristic is always **optimistic**

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Admissible heuristics

The 8-puzzle:

- $h_1(n)$ = number of misplaced tiles
- $h_2(n)$ = total Manhattan distance (i.e., no. of squares from desired location of each tile)

5	4	
6	1	8
7	3	2

1	2	3
8		4
7	6	5

Start State

Goal State

- $h_1(S) = ?$ 7
- $h_2(S) = ?$ $2+3+3+2+4+2+0+2 = 18$

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Heuristic quality

- Effective branching factor b^*
 - Is the branching factor that a uniform tree of depth d would have in order to contain $N+1$ nodes.

$$N + 1 = 1 + b^* + (b^*)^2 + \dots + (b^*)^d$$

- Measure is fairly constant for sufficiently hard problems.
 - Can thus provide a good guide to the heuristic's overall usefulness.
 - A good value of b^* is 1.

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Heuristic quality and dominance

- 1200 random problems with solution lengths from 2 to 24.
- If $h_2(n) \geq h_1(n)$ for all n (both admissible) then h_2 dominates h_1 and is better for search

d	Search Cost			Effective Branching Factor		
	IDS	$A^*(h_1)$	$A^*(h_2)$	IDS	$A^*(h_1)$	$A^*(h_2)$
2	10	6	6	2.45	1.79	1.79
4	112	13	12	2.87	1.48	1.45
6	680	20	18	2.73	1.34	1.30
8	6384	39	25	2.80	1.33	1.24
10	47127	93	39	2.79	1.38	1.22
12	3644035	227	73	2.78	1.42	1.24
14	-	539	113	-	1.44	1.23
16	-	1301	211	-	1.45	1.25
18	-	3056	363	-	1.46	1.26
20	-	7276	676	-	1.47	1.27
22	-	18094	1219	-	1.48	1.28
24	-	39135	1641	-	1.48	1.26

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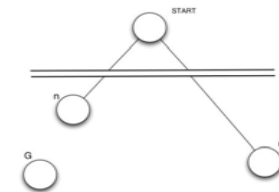
Inventing admissible heuristics

- Admissible heuristics can be derived from the exact solution cost of a relaxed version of the problem:
 - Relaxed 8-puzzle for h_1 : a tile can move anywhere
As a result, $h_1(n)$ gives the shortest solution
 - Relaxed 8-puzzle for h_2 : a tile can move to any adjacent square.
As a result, $h_2(n)$ gives the shortest solution.

The optimal solution cost of a relaxed problem is no greater than the optimal solution cost of the real problem.

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Optimality of A*(standard proof)



- Suppose suboptimal goal G_2 in the queue.
- Let n be an unexpanded node on a shortest to optimal goal G .

$$\begin{aligned}
 f(G_2) &= g(G_2) && \text{since } h(G_2)=0 \\
 &> g(G) && \text{since } G_2 \text{ is suboptimal} \\
 &\geq f(n) && \text{since } h \text{ is admissible}
 \end{aligned}$$

Since $f(G_2) > f(n)$, A* will never select G_2 for expansion

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Optimality for graphs?

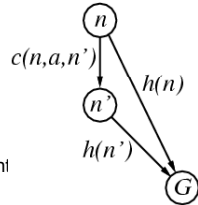
- Admissibility is not sufficient for graph search
 - In graph search, the optimal path to a repeated state could be discarded if it is not the first one generated
 - Can fix problem by requiring consistency property for $h(n)$

- A heuristic is **consistent** if for every successor n' of a node n generated by any action a ,

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

(aka "monotonic")

- admissible heuristics are generally consistent



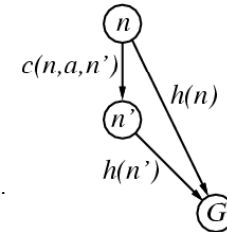
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A* is optimal with consistent heuristics

- If h is consistent, we have

$$\begin{aligned} f(n') &= g(n') + h(n') \\ &= g(n) + c(n,a,n') + h(n') \\ &\geq g(n) + h(n) \\ &= f(n) \end{aligned}$$

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



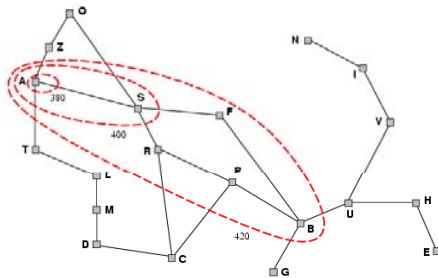
Thus, first goal-state selected for expansion must be optimal

- Theorem:
 - If $h(n)$ is consistent, A* using GRAPH-SEARCH is optimal
 -

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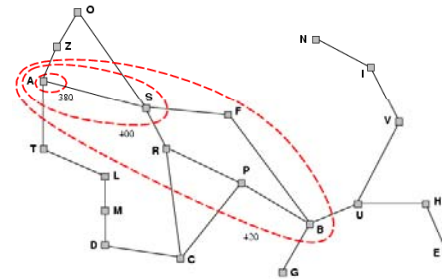
Contours of A* Search

- 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 $f_i < f_{i+1}$



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Contours of A* Search



- With uniform-cost ($h(n) = 0$), contours will be circular
- With good heuristics, contours will be focused around optimal path
- A* will expand all nodes with cost $f(n) < C^*$

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A* search, evaluation

- Completeness: YES
 - Since bands of increasing f are added
 - Unless there are infinitely many nodes with $f < f(G)$

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A* search, evaluation

- Completeness: YES
- Time complexity:
 - Number of nodes expanded is still exponential in the length of the solution.

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A* search, evaluation

- Completeness: YES
- Time complexity: (exponential with path length)
- Space complexity:
 - It keeps all generated nodes in memory
 - Hence space is the major problem not time

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A* search, evaluation

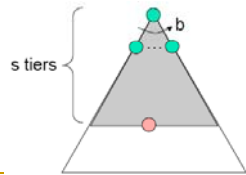
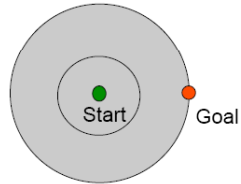
- Completeness: YES
- Time complexity: (exponential with path length)
- Space complexity: (all nodes are stored)
- Optimality: YES
 - Cannot expand f_{i+1} until f_i is finished.
 - A* expands all nodes with $f(n) < C^*$
 - A* expands some nodes with $f(n) = C^*$
 - A* expands no nodes with $f(n) > C^*$

Also *optimally efficient* (not including ties)

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Compare Uniform Cost and A*

- Uniform-cost expanded in all directions



- A* expands mainly toward the goal, but does hedge its bets to ensure optimality

