

Informed search algorithms

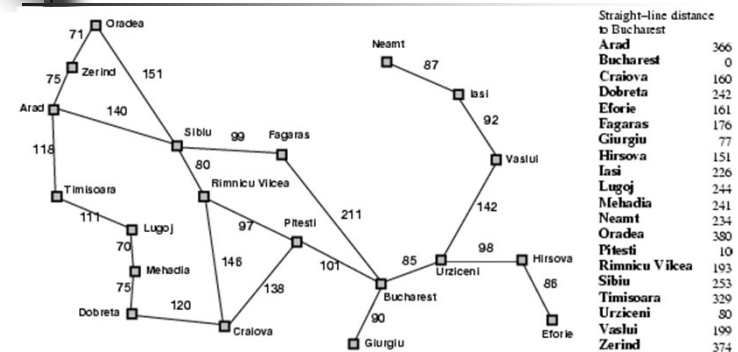
Outline

- Best-first search
- Greedy best-first search
- A* search
- Heuristics
- Local search algorithms
- Hill-climbing search
- Simulated annealing search
- Local beam search
- Genetic algorithms

Best-first search

- Idea: use an evaluation function $f(n)$ for each node
 - $f(n)$ provides an estimate for the total cost.
 - Expand the node n with smallest $f(n)$.
- Implementation:
Order the nodes in fringe increasing order of cost.
- Special cases:
 - greedy best-first search
 - A* search

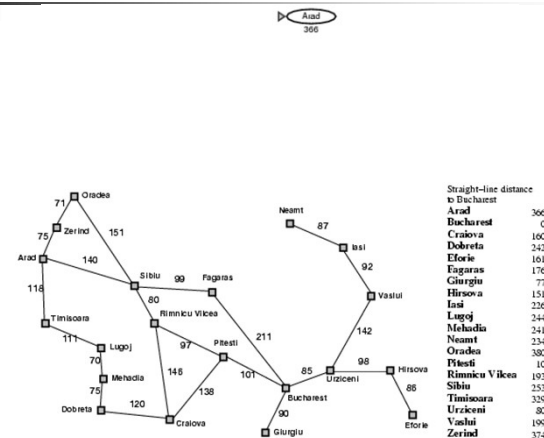
Romania with straight-line dist.



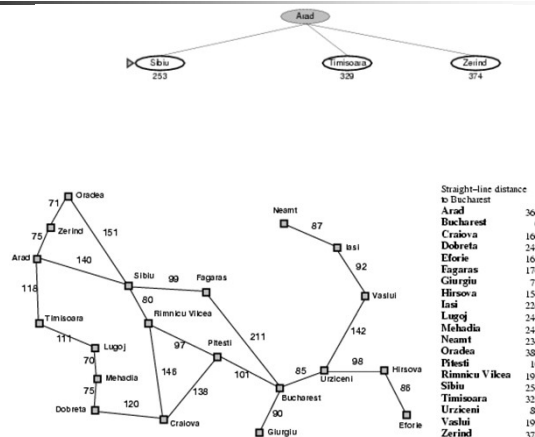
Greedy best-first search

- $f(n)$ = estimate of cost from n to *goal*
- e.g., $f_{SLD}(n)$ = straight-line distance from n to Bucharest
- Greedy best-first search expands the node that appears to be closest to goal.

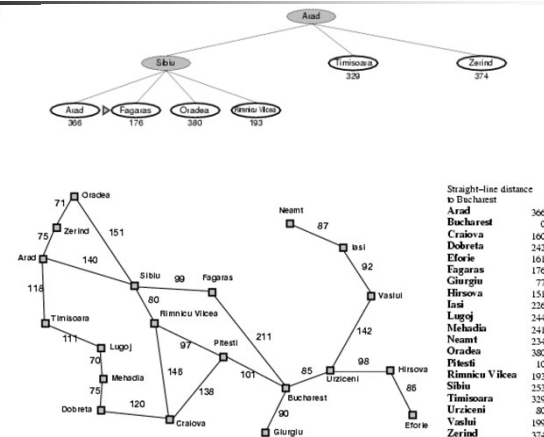
Greedy best-first search example



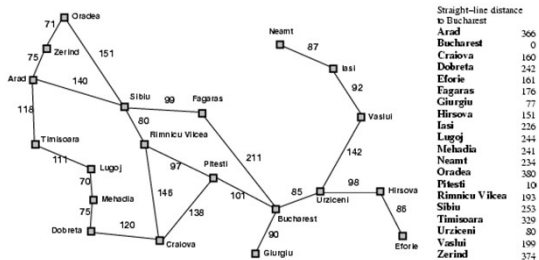
Greedy best-first search example



Greedy best-first search example



Greedy best-first search example



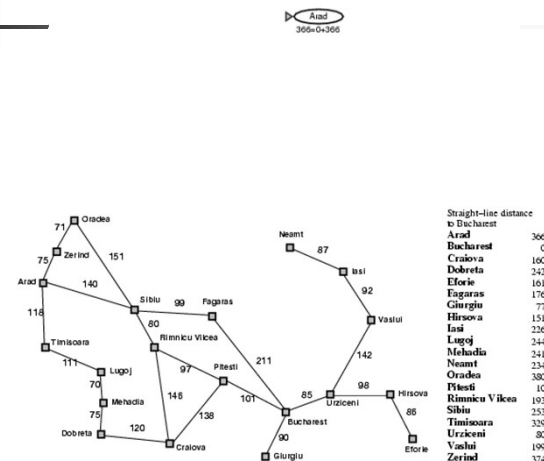
Properties of greedy best-first search

- Complete? No – can get stuck in loops.
- Time? $O(b^m)$, but a good heuristic can give dramatic improvement
- Space? $O(b^m)$ - keeps all nodes in memory
- Optimal? No
e.g. Arad → Sibiu → Rimnicu Virea → Pitesti → Bucharest is shorter!

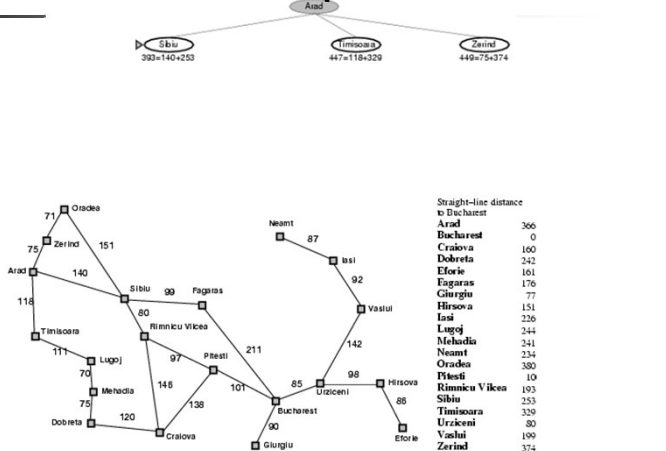
A* search

- Idea: avoid expanding paths that are already expensive
- 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
- Best First search has $f(n) = h(n)$

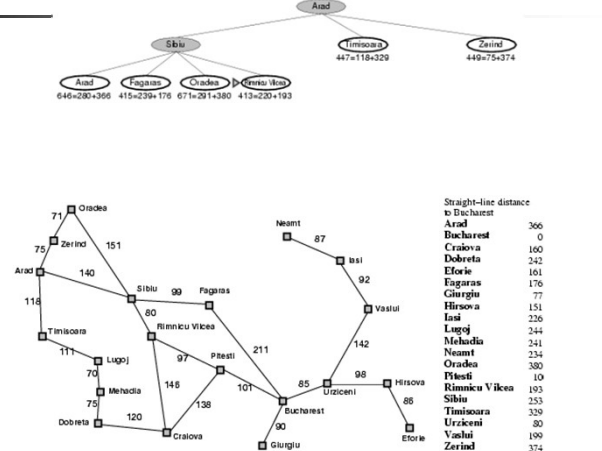
A* search example



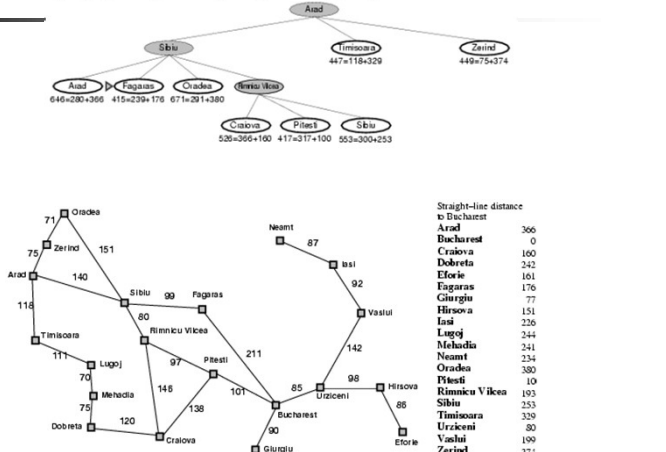
A* search example



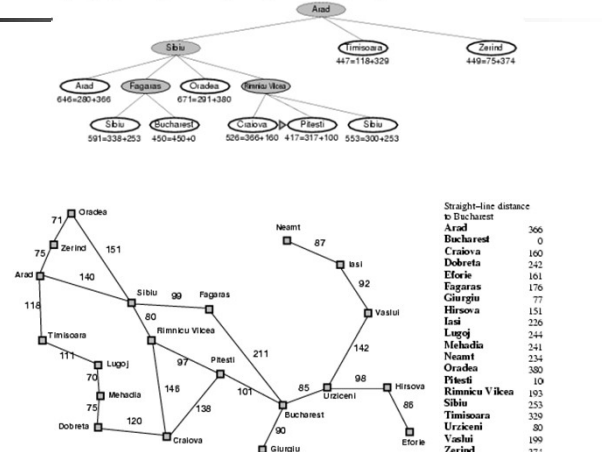
A* search example



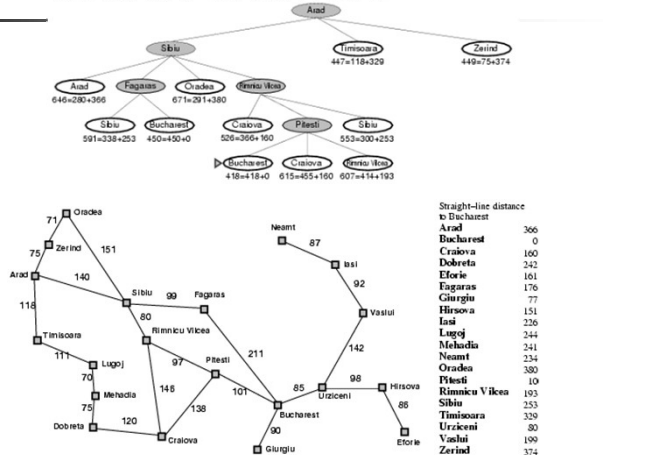
A* search example



A* search example



A* search example



Admissible heuristics

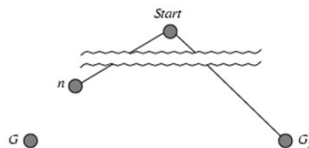
- A heuristic $h(n)$ is admissible if for every node n , $h(n) \leq h^*(n)$, where $h^*(n)$ is the true cost to reach the goal state from n .
- An admissible heuristic never overestimates the cost to reach the goal, i.e., it is optimistic
- Example: $h_{SLD}(n)$ (never overestimates the actual road distance)
- Theorem: If $h(n)$ is admissible, A* using TREE-SEARCH is optimal

Optimality of A* (proof)

- Suppose some suboptimal goal G_2 has been generated and is in the fringe. Let n be an unexpanded node in the fringe such that n is on a shortest path to an optimal goal G .

We want to prove:
 $f(n) < f(G_2)$
 (then A* will prefer n over G_2)

- $f(G_2) = g(G_2)$ since $h(G_2) = 0$
- $f(G) = g(G)$ since $h(G) = 0$
- $g(G_2) > g(G)$ since G_2 is suboptimal
- $f(G_2) > f(G)$ from above

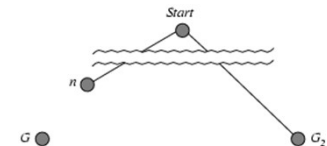


Optimality of A* (proof)

- Suppose some suboptimal goal G_2 has been generated and is in the fringe. Let n be an unexpanded node in the fringe such that n is on a shortest path to an optimal goal G .

- $f(G_2) > f(G)$ copied from last slide
- $h(n) \leq h^*(n)$ since h is admissible (*under-estimate*)
- $g(n) + h(n) \leq g(n) + h^*(n)$ from above
- $f(n) \leq f(G)$ since $g(n) + h(n) = f(n)$ & $g(n) + h^*(n) = f(G)$
- $f(n) < f(G_2)$ from top line.

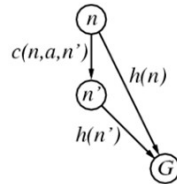
Hence: n is preferred over G_2



Consistent heuristics

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

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



It's the triangle inequality!

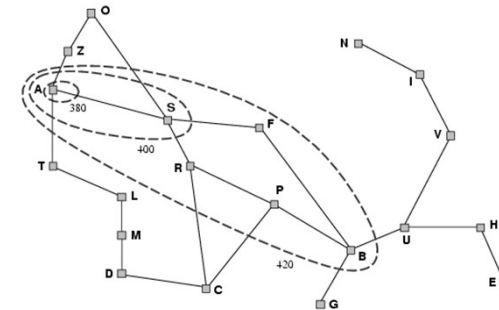
- If h is consistent, we have

$$\begin{aligned} f(n') &= g(n') + h(n') \\ &= g(n) + c(n,a,n') + h(n') \\ &\geq g(n) + h(n) = f(n) \\ f(n') &\geq f(n) \end{aligned}$$
- i.e., $f(n)$ is non-decreasing along any path.
- Theorem: If $h(n)$ is consistent, A* using GRAPH-SEARCH is optimal

keeps all checked nodes
in memory to avoid repeated
states

Optimality of A*

- A* expands nodes in order of increasing f value
- Gradually adds " f -contours" of nodes
- Contour i contains all nodes with $f \leq f_i$ where $f_i < f_{i+1}$



Properties of A*

- Complete? Yes (unless there are infinitely many nodes with $f \leq f(G)$, i.e. path-cost $> \epsilon$)
- Time/Space? Exponential b^d
except if: $|h(n) - h^*(n)| \leq O(\log h^*(n))$
- Optimal? Yes
- Optimally Efficient? Yes (no algorithm with the same heuristic is guaranteed to expand fewer nodes)

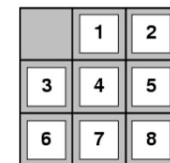
Admissible heuristics

E.g., for 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)



Start State



Goal State

- $h_1(S) = ?$
- $h_2(S) = ?$

Admissible heuristics

E.g., for the 8-puzzle:

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

| | | |
|---|---|---|
| 7 | 2 | 4 |
| 5 | | 6 |
| 8 | 3 | 1 |

Start State

| | | |
|---|---|---|
| | 1 | 2 |
| 3 | 4 | 5 |
| 6 | 7 | 8 |

Goal State

- $h_1(S) = ?$ 8
- $h_2(S) = ?$ $3+1+2+2+2+3+3+2 = 18$