

Informed search

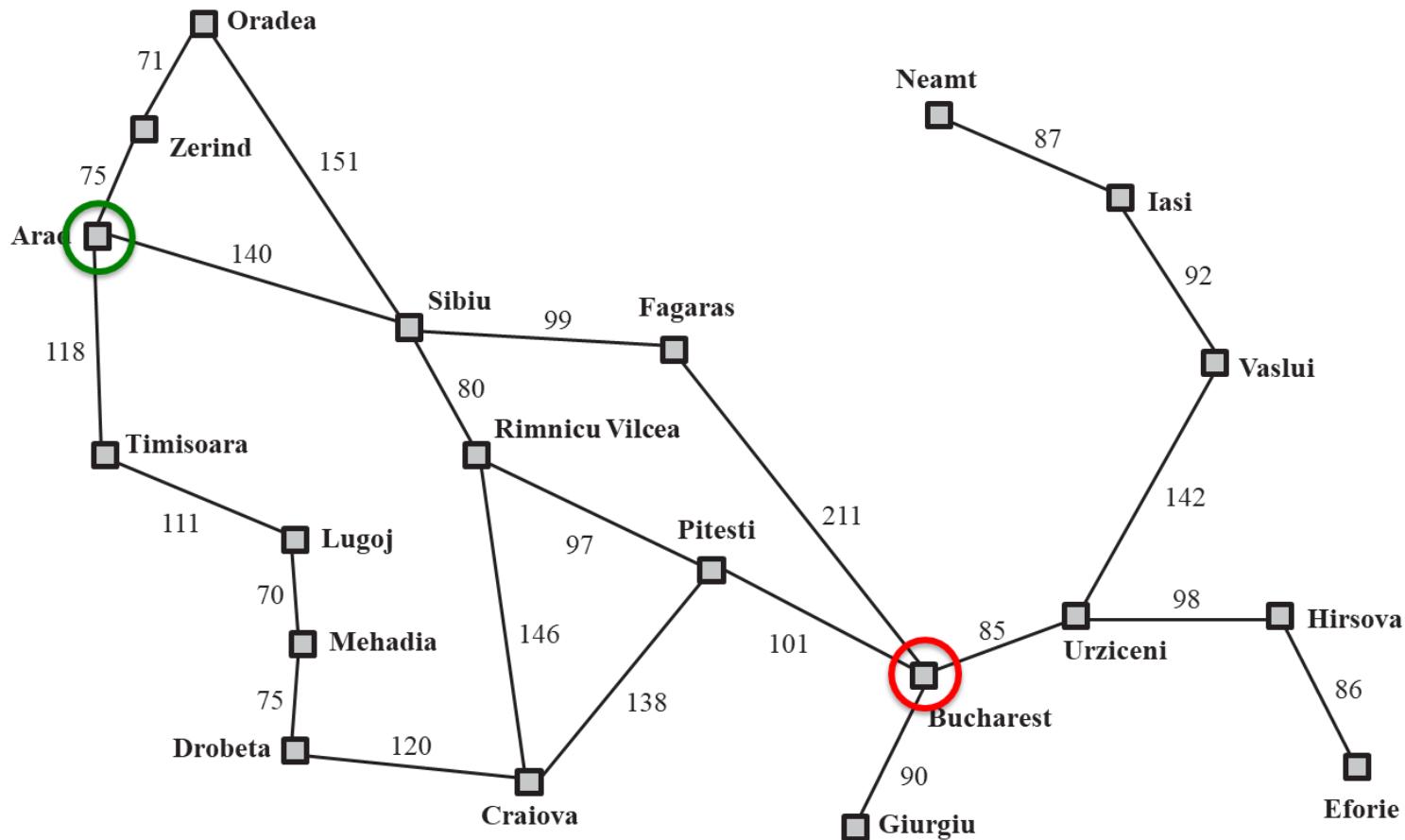
EMI | Semestre 1 | Pr Mohamed RHAZZAF

Recap on search

- Search problem:
 - States (configurations of the world)
 - Actions and costs
 - Successor function (world dynamics)
 - Start state and goal test
- Search tree:
 - Nodes: represent plans for reaching states
 - Plans have costs (sum of action costs)
- Search algorithm:
 - Systematically builds a search tree
 - Chooses an ordering of the fringe (unexplored nodes)
 - Optimal: finds least-cost plans



Example: route-finding in Romania

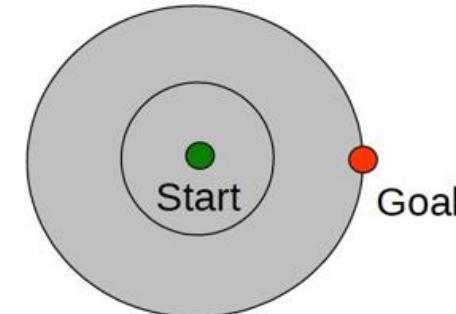
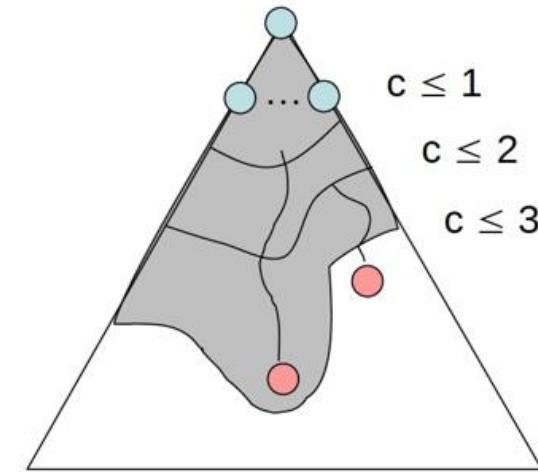


Search Strategies

- Systematic Non-informed search: Find a path, when exploring space, that leads from the initial state to the final state
 - depth first and breath first
- **Search with information about edge (special case)**
 - uniform cost search (dijkstra)
- **Search with information about node**
 - Hill-Climbing, Best-First and Beam
- **Search with information about nodes and edges**
 - A*

Uniform Cost Search

- Strategy : expend lowest path cost from the start
- The good thing: UCS is complete and optimal
- The bad:
 - Explores options in every “direction”
 - No information about goal location



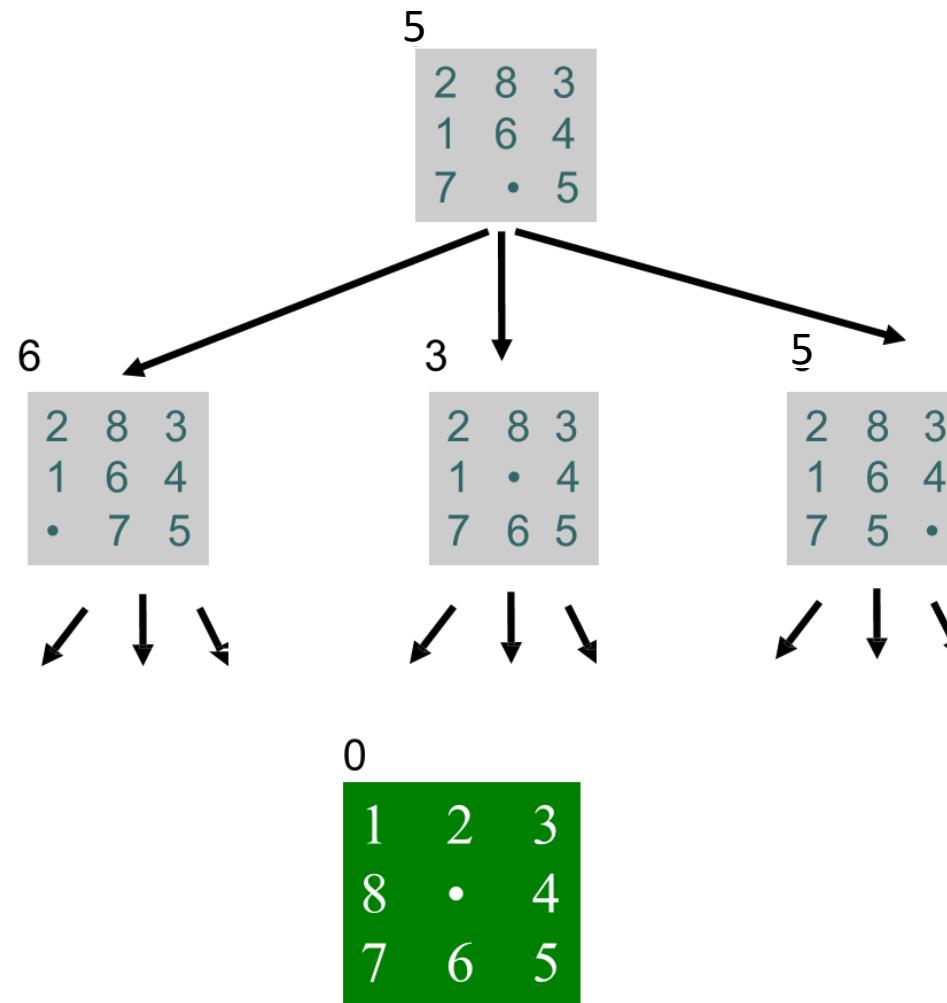
Search Strategies

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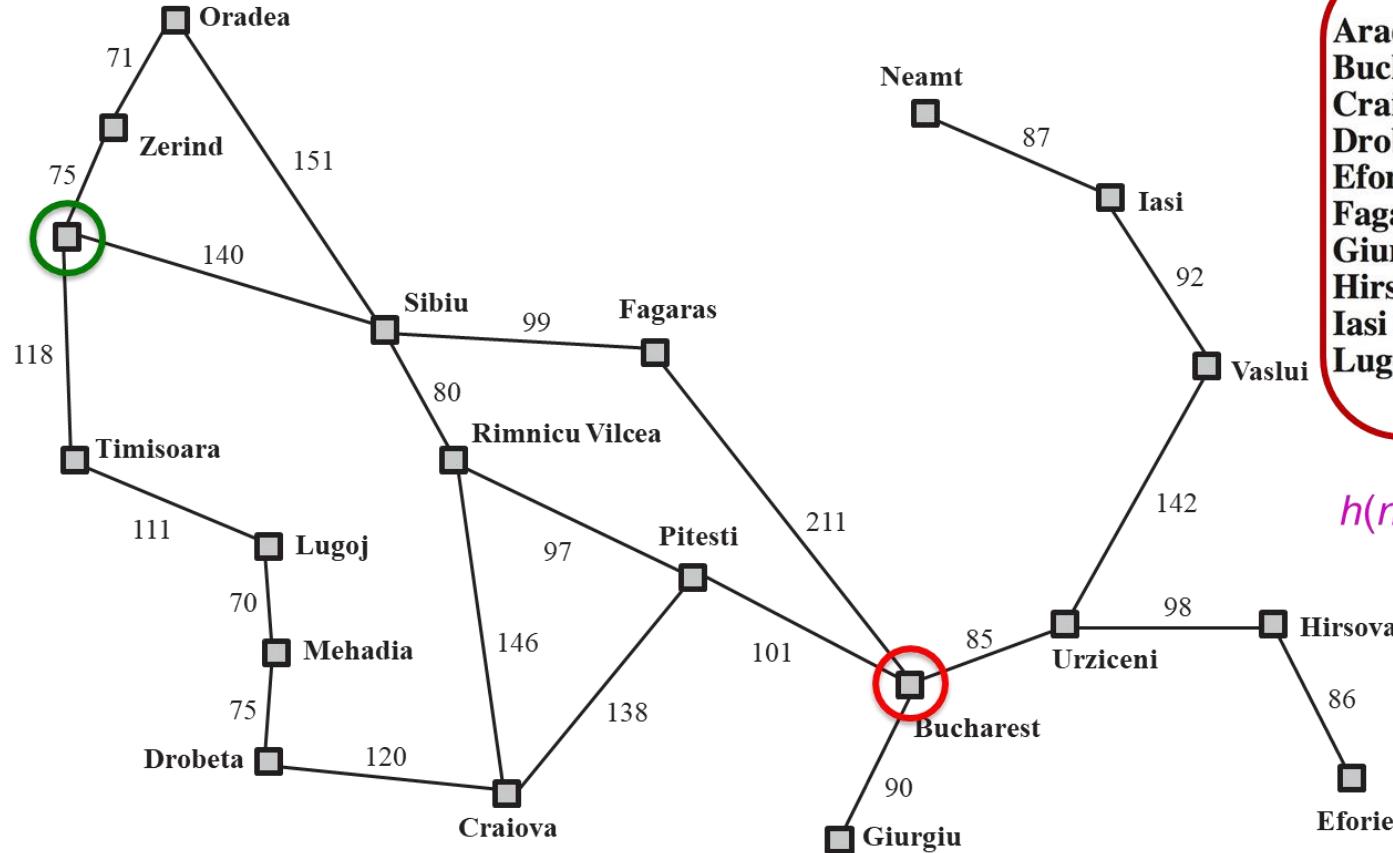
Heuristic method

- **Systematic search**
 - Explore all paths
 - Eventually find a solution
 - Eventually may take a long time
- **Heuristic search**
 - Use of heuristics: choose paths to explore in priority
 - Heuristic criterion: associate with each state a certain estimated value (its approximation of the goal)
 - Ex: $h(n)$ = the distance between city n and destination city
 - For each problem, different estimate
 - Heuristic search: first explore the successor state representing the lowest cost
 - Efficiency of heuristic research depends on how the knowledge of the field is exploited

Example: 8 puzzle heuristic



Example: route-finding in Romania

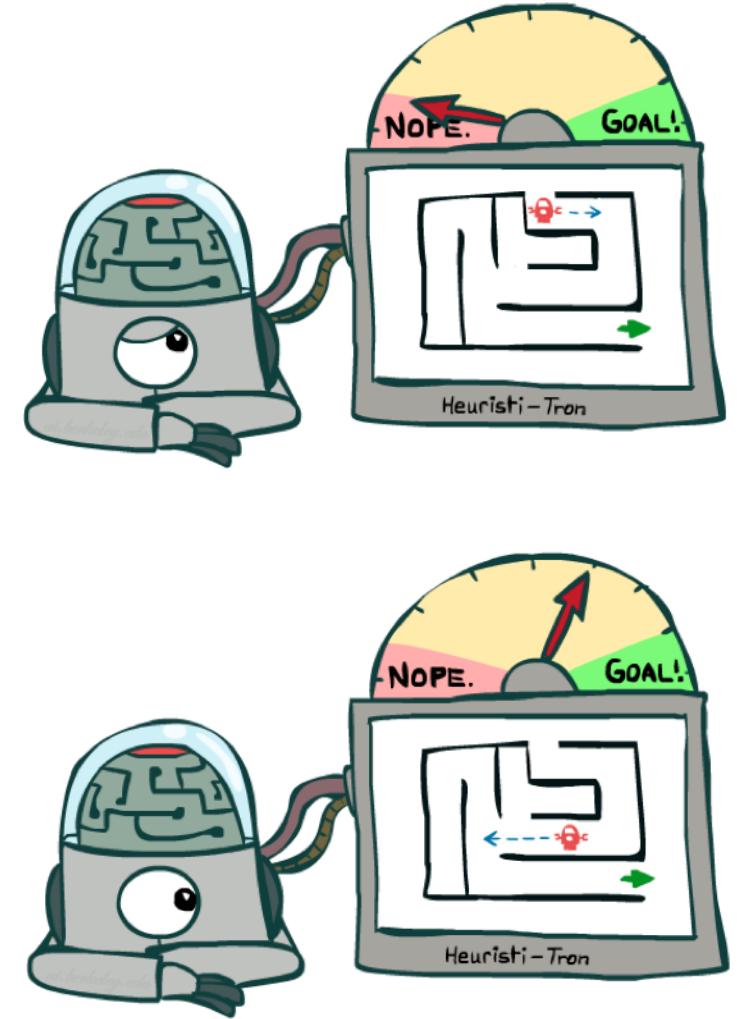
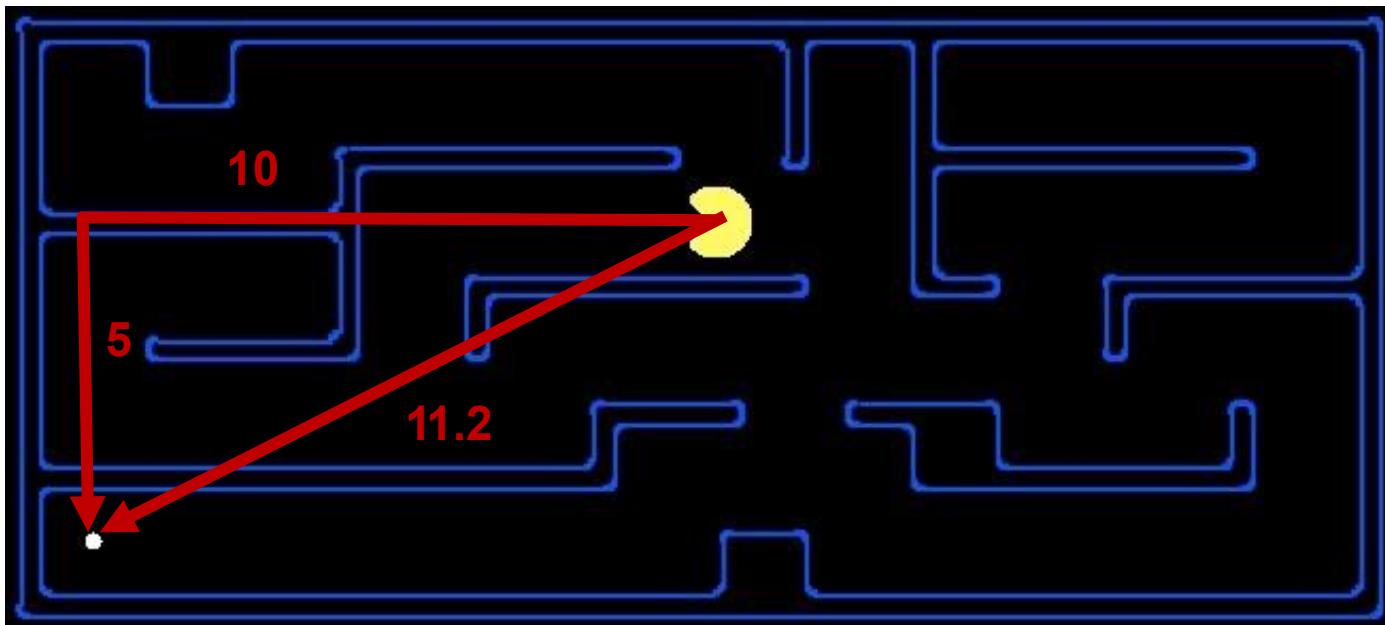


Arad	366	Mehadia	241
Bucharest	0	Neamt	234
Craiova	160	Oradea	380
Drobeta	242	Pitesti	100
Eforie	161	Rimnicu Vilcea	193
Fagaras	176	Sibiu	253
Giurgiu	77	Timisoara	329
Hirsova	151	Urziceni	80
Iasi	226	Vaslui	199
Lugoj	244	Zerind	374

$h(n)$ = straight-line distance to Bucharest

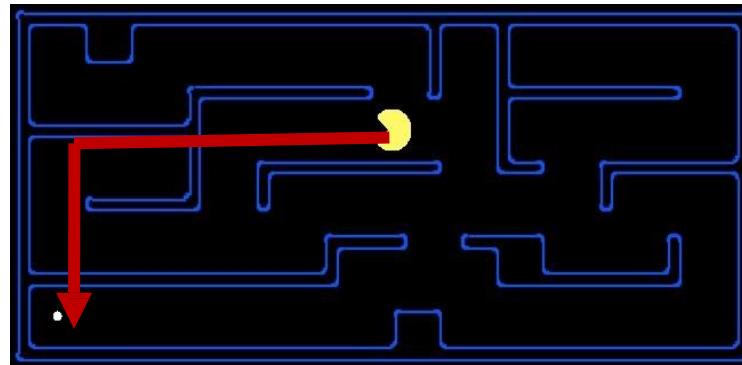
Search Problems

- $h(n) = \text{Manhattan distance} = |x| + |y|$
- Is Manhattan better than straight-line distance?



Admissible Heuristics

- A heuristic h is admissible (optimistic) if:
 - $0 \leq h(n) \leq h^*(n)$ where $h^*(n)$ is the true cost to a nearest goal
- Example:



- Finding good, cheap admissible heuristics is the key to success

Hill Climbing

- **Simple, general idea**

- **Initialization:** Start with a random or arbitrary initial solution.
- **Evaluation:** Evaluate the quality of the current solution using a fitness or objective function.
- **Neighbor generation:** Generate neighboring solutions by making small changes to the current one.
- **Neighbor selection:** Compare the neighbors to the current solution. If a neighbor is better, choose it as the new current solution.
- **Iteration:** Repeat the evaluation and neighbor selection steps until no neighboring solution is better than the current one.
- **Termination:** The algorithm stops when it reaches a point where it cannot improve further, meaning it has found a local optimum



Hill-climbing algorithm

```
function HILL-CLIMBING(problem) returns a state
```

```
    current ← make-node(initial-state)
```

```
    loop do
```

```
        neighbor ← a highest-valued successor of current
```

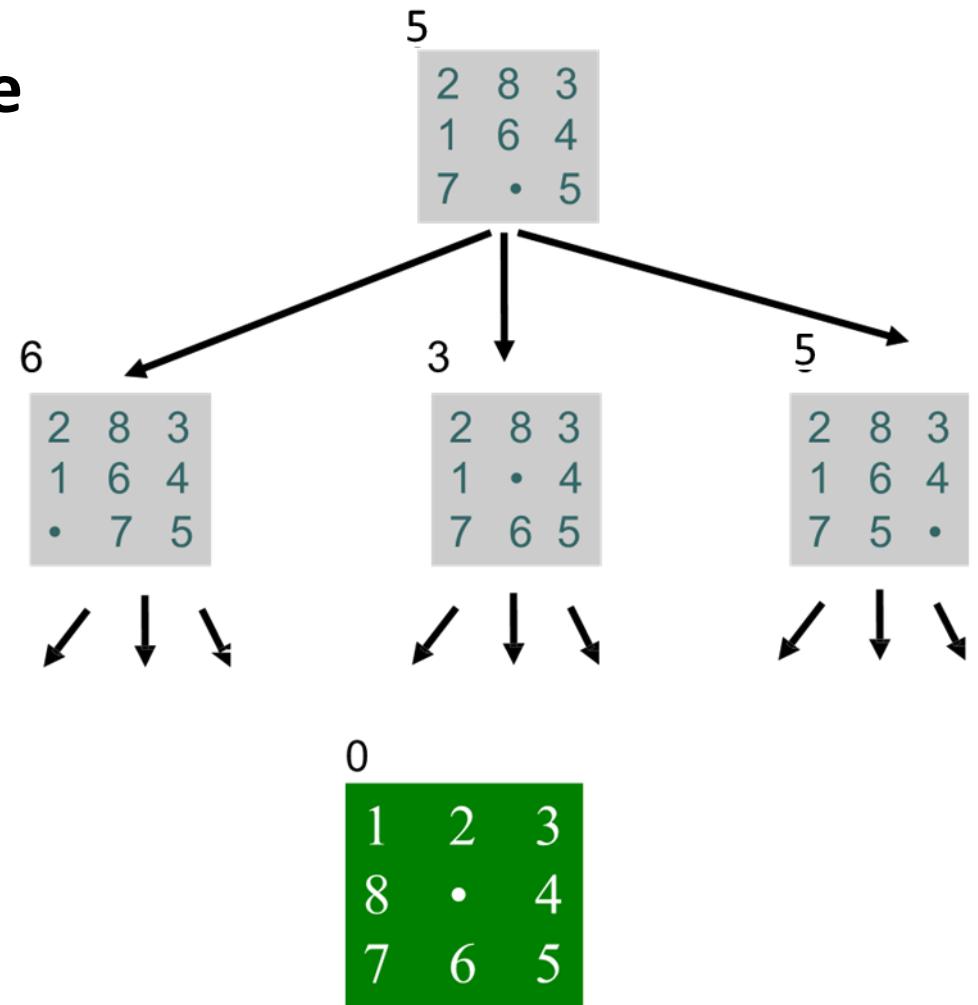
```
        if neighbor.value ≤ current.value then
```

```
            return current.state
```

```
        current ← neighbor
```

Examples

8 puzzle

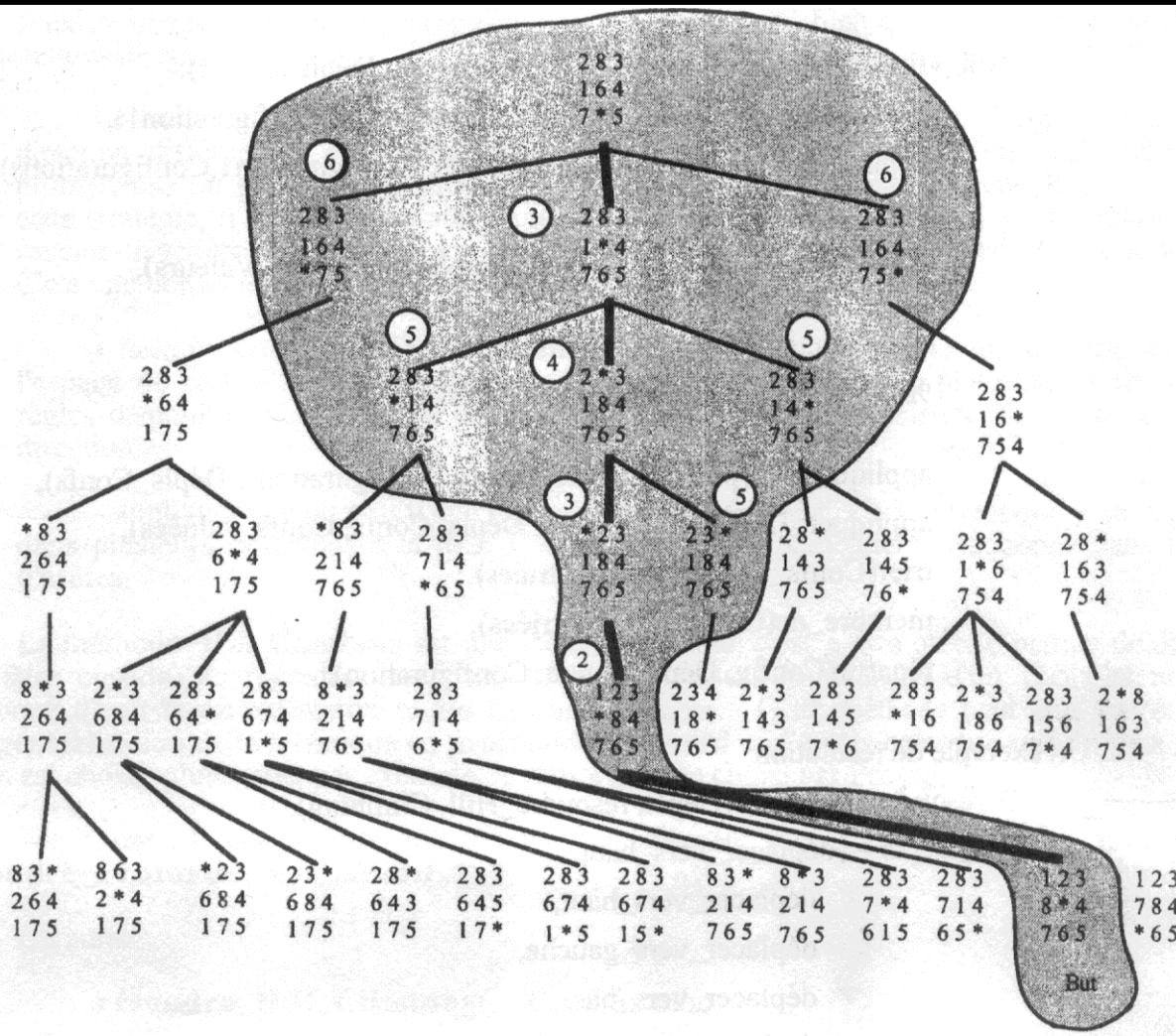


Hill Climbing Variant

- **Problem with simple Hill climbing**
 - The algorithm considers only one path
when the successors are not better than this state, the algorithm stops
- **Hill climbing v2**

```
function HILL-CLIMBING_V2(problem) returns a state
    current ← make-node(initial-state)
    loop do
        neighbor ← a highest-valued successor of current
        —if neighbor.value ≤ current.value then—
            return current.state
        current ← neighbor
```

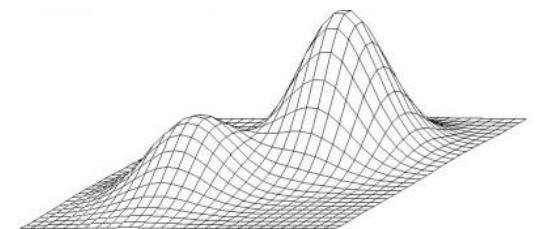
Hill Climbing Variant



Hill Climbing

- **Key characteristics**

- **Local search:** It only explores the neighborhood of the current solution, not the entire search space.
- **Greedy:** It makes the best immediate move at each step without considering future consequences.
- **No backtracking:** It cannot go back to previous states, meaning it can get stuck in local optima



- **Hill Climbing**

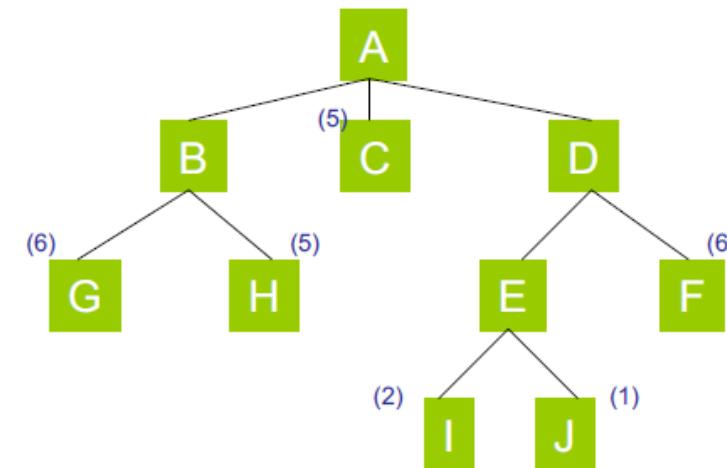
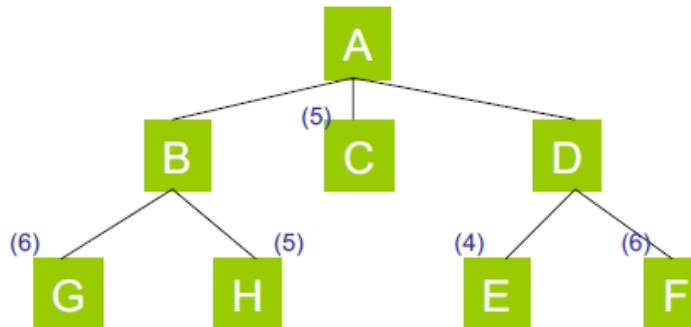
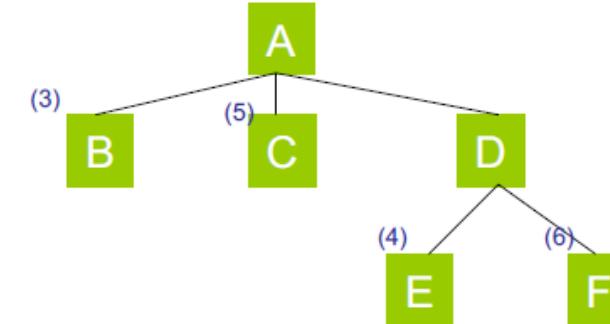
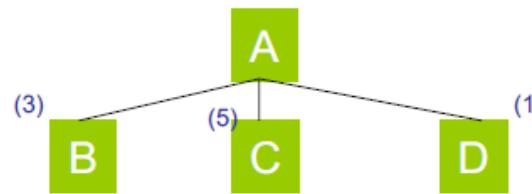
- Local technique, i.e. considers only the immediate consequences of a state

Best First

- **Best First**
 - Combination of depth and breath search
 - Follow one path at a time but change it as soon as a more promising path appears
- **Method**
 - Select the most promising node of all generated nodes
 - Perform the expansion of the node using the applicable rules
 - Among the successors, if one of them is a solution, stop
 - Otherwise, all new nodes are added to all the nodes already generated
 - Again, the most promising node becomes the current node

Example Best First

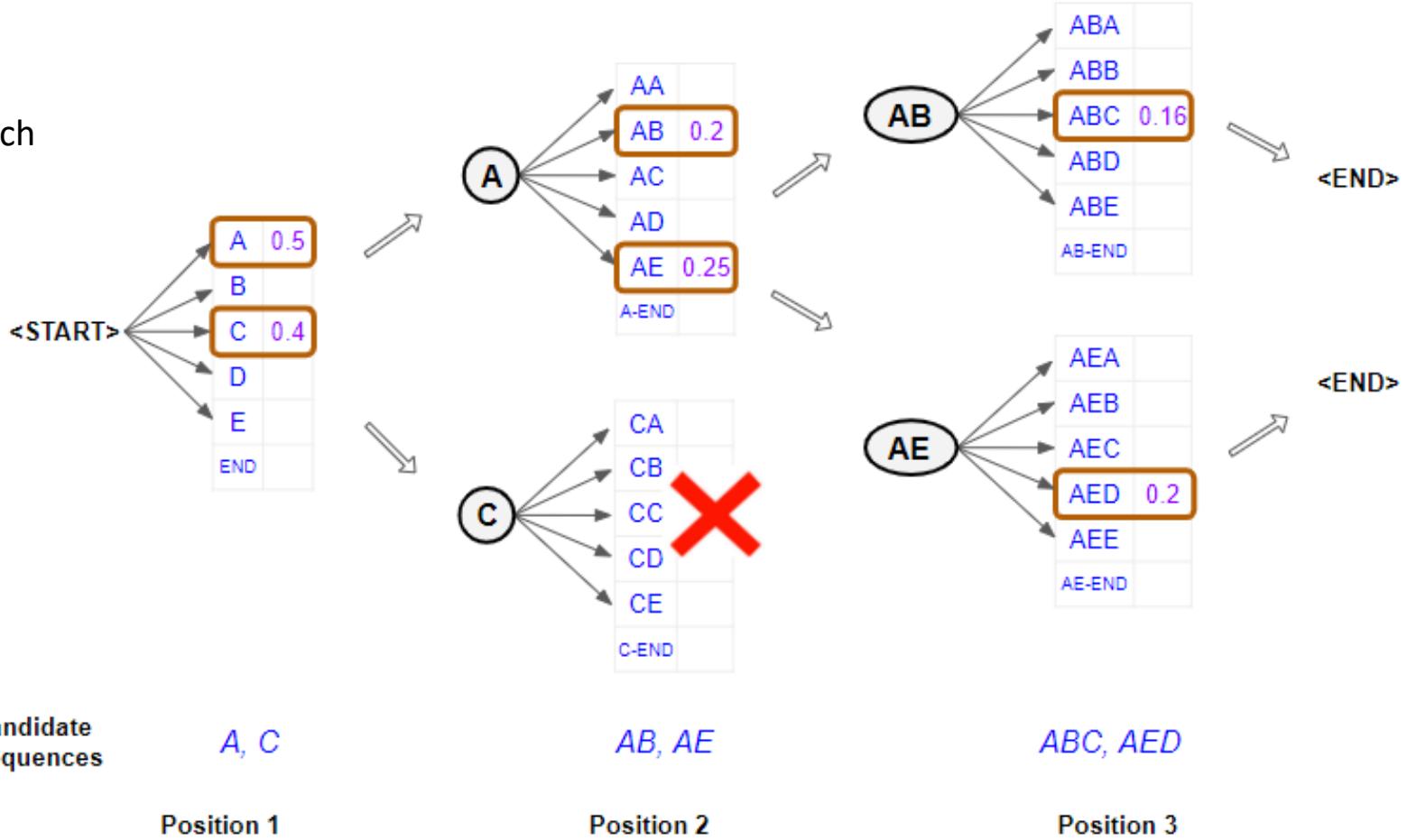
A



Beam search

- Beam search

- Is an optimization of best-first search
- Reduces memory load
- In the beam search method, only a certain number **K** of solutions are kept as candidates



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Generic definition of f

- In practice we do not know the distance to the goal! That's what we're looking for
- On the other hand, we know the optimal distance in the explored part between the root and a node already explored
- It is convenient to separate $f(n)$ into two parts:
 - $g(n)$: the real cost of the optimal path from the root to n in the part already explored
 - $h(n)$: estimated cost of the rest of the path from n to the goal. $h(n)$ is called the heuristic function.

A* Search



Generic definition of f

- Depending on the weight we want to give to either party, we define f as follows

$$f(n) = (1-w) \cdot g(n) + w \cdot h(n)$$

where w is a real number greater than or equal to 0 and less than or equal to 1

- Depending on the values given to w, one obtains classical search algorithms:
 - Dijkstra : $w = 0$, ($f(n) = g(n)$)
 - Best-first search : $w = 1$, ($f(n) = h(n)$)
 - A* : $w = 0.5$, ($f(n) = g(n) + h(n)$) e.g Information about nodes and edges

A* Example

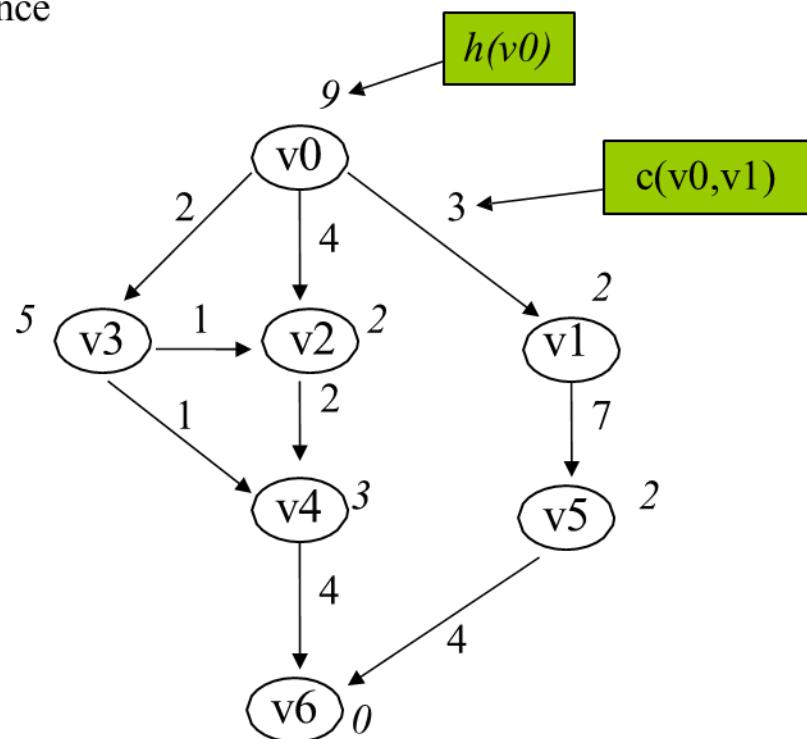
Distance between cities:

v0: departure

v6: destination

h: distance

C: actual distance



Exploration:

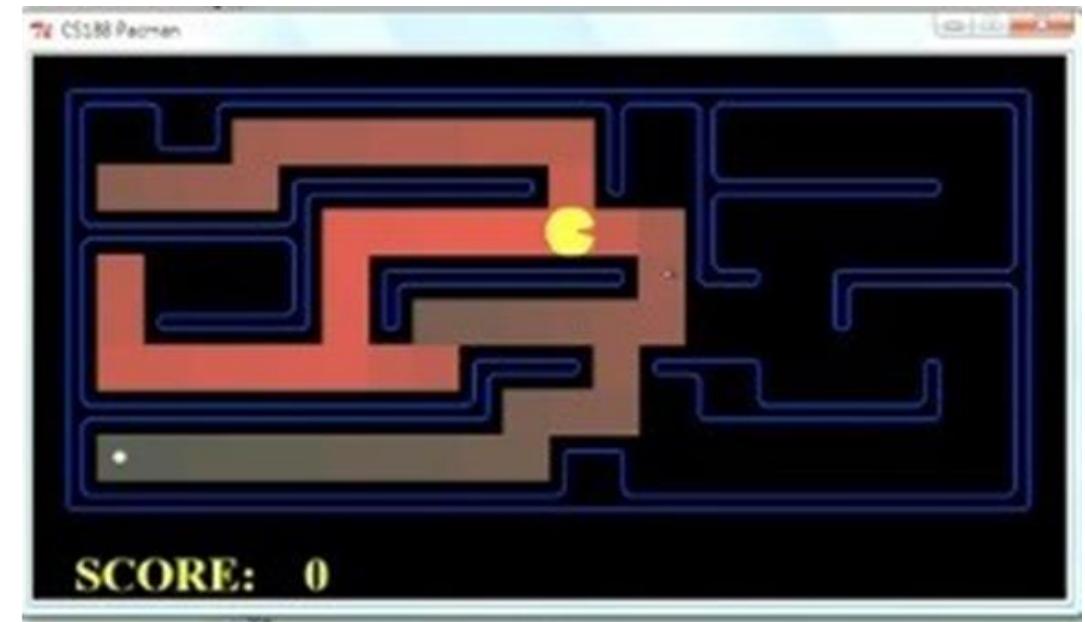
1. (v0, 9, void)
2. (v1, 5, v0) (v2, 6, v0), (v3, 7, v0)
3. (v2, 6, v0) (v3, 7, v0), (v5, 12, v1)
4. (v3, 7, v0), (v4, 9, v2), (v5, 12, v1)
5. (v4, 6, v3), (v4, 9, v2), (v5, 12, v1)

Solution: **v0,v3,v4,v6**

Comparison



Uniform Cost (g)



A* (g+h)

A* Applications

- Video games
 - Pathing / routing problems
 - Resource planning problems
 - Robot motion planning
 - Language analysis
 - Machine translation
 - Speech recognition
 - Protein design
 - Chemical synthesis
 - ...

