UNINFORMED SEARCH (FINALE) + INFORMED SEARCH

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> 9/14/2023 CMSC 671

By the end of class today, you will be able to:

- 1. Assess the limitation of uninformed search
- 2. Select the right uninformed search algorithm for a given task.

Modified from slides by Dr. Cassandra Kent

RECAP

BFS VS DFS

Breadth-First Search

- Complete
- Optimal
- $O(b^d)$ time complexity
- $O(b^d)$ space complexity

Depth-First Search

- Not complete
- Not optimal
- $O(b^m)$ time complexity
- O(bm) space complexity

RECAP

BFS VS UCS

Breadth-First Search

- Complete
- Optimal
- $O(b^d)$ time complexity
- $O(b^d)$ space complexity

Uniform-Cost Search

- Complete
- Optimal
- $O(b^{1+C^*/\epsilon})$ time complexity
- $O(b^{1+C^*/\epsilon})$ space complexity

DEPTH-LIMITED SEARCH

PROPERTIES OF DEPTH-LIMITED SEARCH

- Is depth-limited search complete?
- Is depth-limited search optimal?

Take a couple minutes and try to figure out under what conditions depth-limited search is complete or optimal.

Some terms that may help: b: branching factor l: depth limit

d: depth of the goal *m*: maximum depth of tree

PROPERTIES OF DEPTH-LIMITED SEARCH

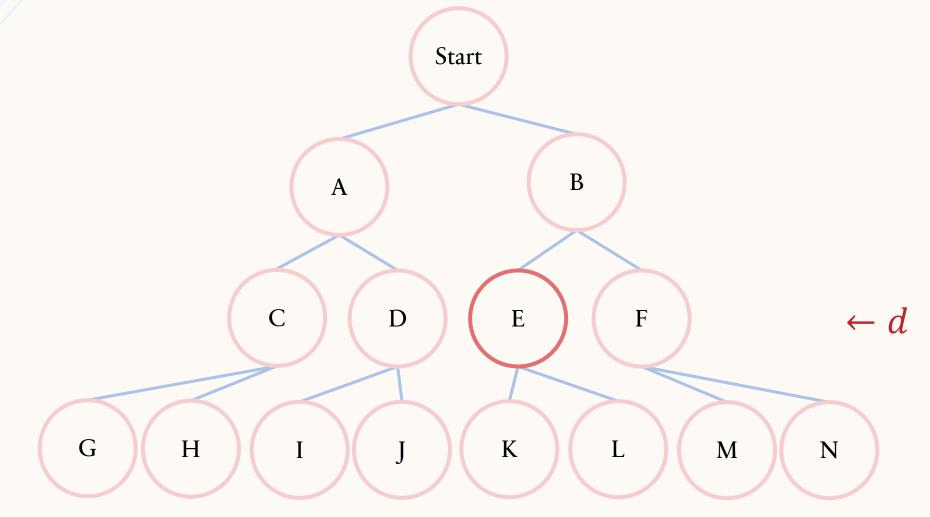
When is depth-limited search complete?

• If the depth limit is **deeper than** the solution depth $(l \ge d)$

b: branching factor *l*: depth limit

d: depth of the goal *m*: maximum depth of tree

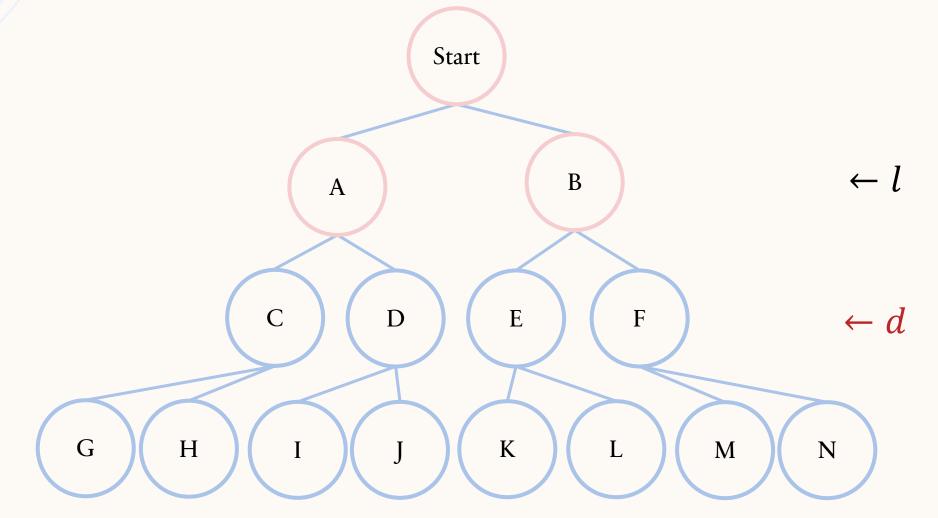
PROPERTIES OF DEPTH-LIMITED SEARCH



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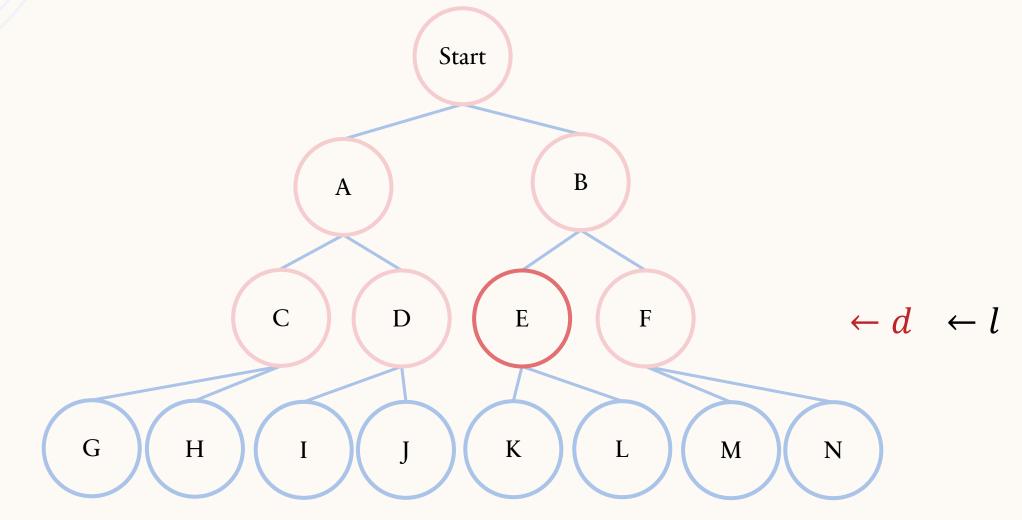
PROPERTIES OF DEPTH-LIMITED SEARCH



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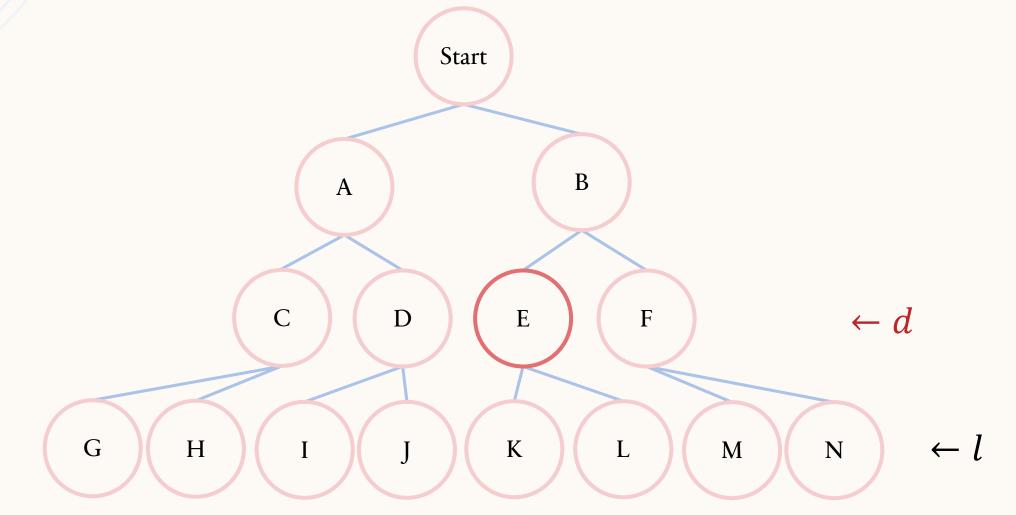
PROPERTIES OF DEPTH-LIMITED SEARCH



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PROPERTIES OF DEPTH-LIMITED SEARCH



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PROPERTIES OF DEPTH-LIMITED SEARCH

When is depth-limited search complete?

• If the depth limit is **deeper than** the solution depth $(l \ge d)$

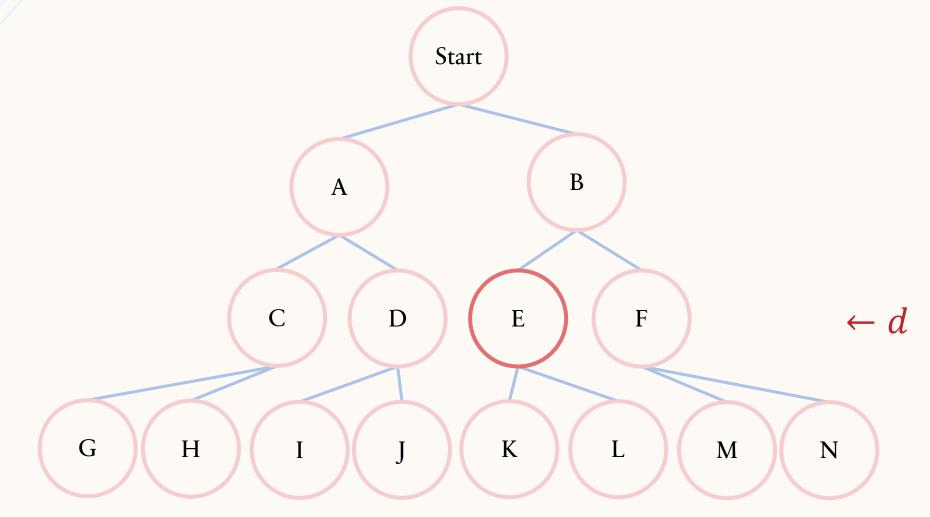
When is depth-limited search optimal?

• If the depth limit is at the optimal solution depth (l = d)

b: branching factor *l*: depth limit

d: depth of the goal *m*: maximum depth of tree

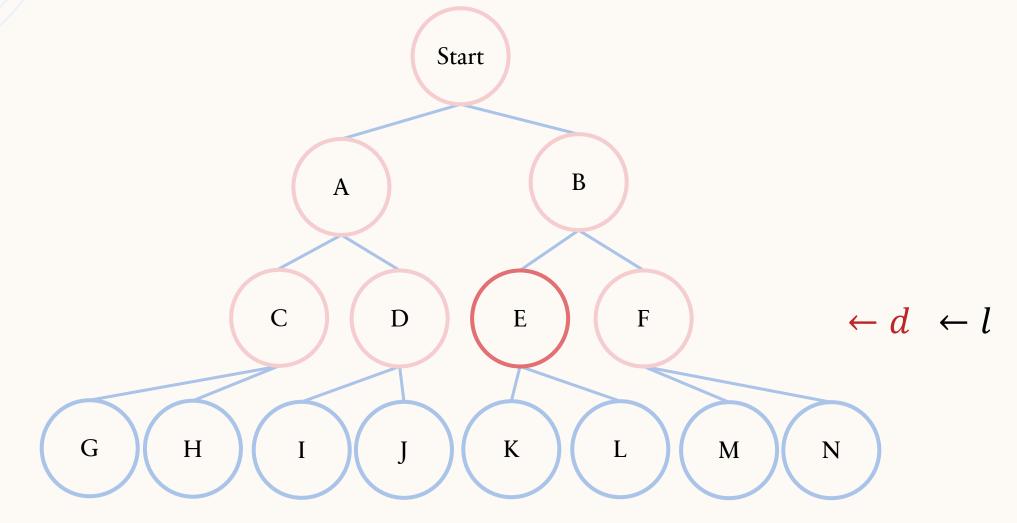
PROPERTIES OF DEPTH-LIMITED SEARCH



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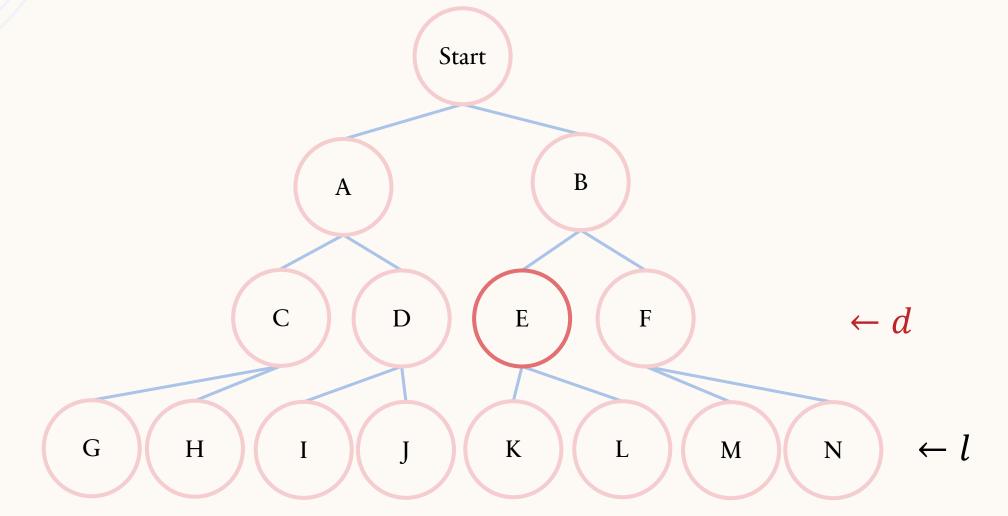
PROPERTIES OF DEPTH-LIMITED SEARCH



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PROPERTIES OF DEPTH-LIMITED SEARCH



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PROPERTIES OF DEPTH-LIMITED SEARCH

We now have a **complete**, **optimal** version of DFS... ...as long as we know the optimal solution depth.

For most problems, we can not determine the optimal solution depth without (optimally) solving the problem first!

Using Depth-Limited Search, can we ensure that when we *first* reach a goal state, it's at the *lowest possible depth*?

ITERATIVE DEEPENING DEPTH-FIRST SEARCH

Brute-force approach to depth limit selection: run depth-limited search with steadily increasing depth limits until a solution is found

function Iterative-Deepening-Search(problem)

```
for depth = 0 to \infty do
```

if result ≠ failure then return result

BFS VS DFS: COMPLETENESS

Problem: Brute-force simple password guessing agent
Depth-Limited Search Breadth

- States expanded (iteration 1): a, b, c
- States expanded (iteration 2): a, aa, ab, ac, ..., b, ba, bb, bc, ...
- States expanded (iteration 2): a, aa, aaa, aab, aac, ..., b, ba, baa, bab, bac, ...

Breadth-First Search

• States expanded: a, b, c, ..., aa, ab, ac, ..., ba, bb, bc, ..., ca, cb, ..., aaa, aab, aac, ... baa, bab, bac, ...

Similar behavior, expands all nodes at lowest level before continuing

ITERATIVE DEEPENING DFS: TIME COMPLEXITY

Compared to BFS, Iterative Deepening re-generates l - 1 nodes at every iteration. Does this affect the time complexity?

We can answer this question using knowledge from Algorithms, specifically Big-O Notation!

ITERATIVE DEEPENING DFS: TIME COMPLEXITY

 $(d)b^{1} + (d-1)b^{2} + (d-2)b^{3} + \dots + (1)b^{d}$

How many nodes does iterative deepening generate in the worst case?

Generate first layer of nodes once for every iteration Generate second layer of nodes once for every iteration, after the first Generate first layer of nodes once for every iteration, after the second Generate final layer of nodes once

ITERATIVE DEEPENING DFS: TIME COMPLEXITY

How many nodes does iterative deepening generate in the worst case?

$$(d)b^{1} + (d-1)b^{2} + (d-2)b^{3} + \dots + (1)b^{d}$$

Asymptotically negligible $= O(b^d)$

Equivalent to BFS!

ITERATIVE DEEPENING PROPERTIES

Breadth-First Search

- Complete
- Optimal
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- Not complete
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Iterative Deepening

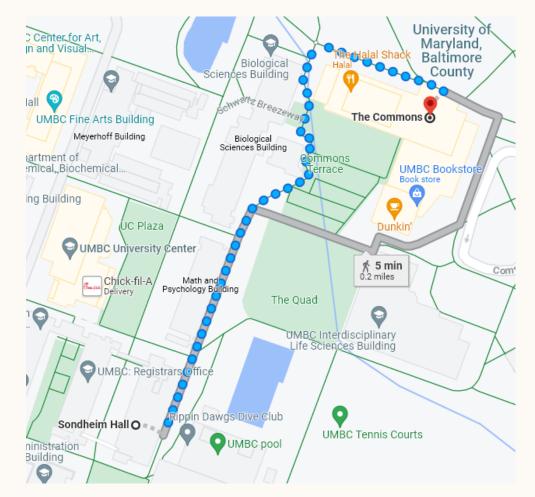
- Complete
- Optimal
- $O(b^d)$ time complexity
- O(bd) space complexity

Finally, a complete, optimal, and memory efficient algorithm! Iterative Deepening DFS is a standard uninformed search method for large search spaces with unknown solution depth.

(i.e. many problems where we would use search)

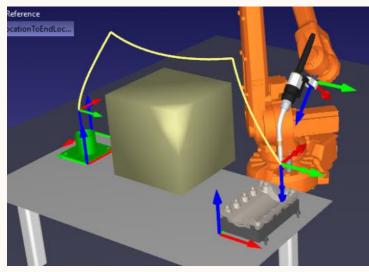
ROUTE FINDING

- State: agent location (which path intersection are you at)
- Actions: follow path *x* to the next intersection
- Goal: find the shortest path to the end location
- Algorithm : Uniform-Cost Search, because we're optimizing for route distance instead of minimum number of actions



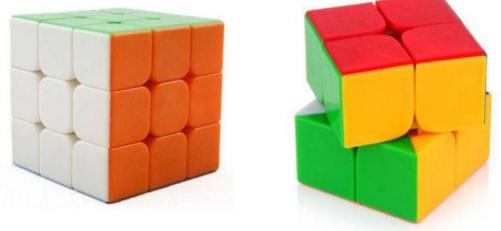
MOTION PLANNING

State: location of obstacles in environment, joint angles of robot
Actions: rotate joints 0, 1, 2, ... to angles i, j, k, ...
Goal: move to the goal configuration over the shortest distance
Algorithm : Uniform-Cost Search, because we're optimizing for the distance the robot moves instead of minimum number of actions



RUBIK'S CUBE

State: configuration of blocks on the cube Actions: rotate each face clockwise/counterclockwise 90 degrees Goal: reach solved state in the fewest moves **Algorithm** : Iterative Deepening DFS, because we're optimizing for fewest moves in a large search space (i.e. memory efficiency matters), and we don't know the optimal solution depth



CHESS

State: configuration of pieces on the board Actions: move 1 piece according to the rules of chess

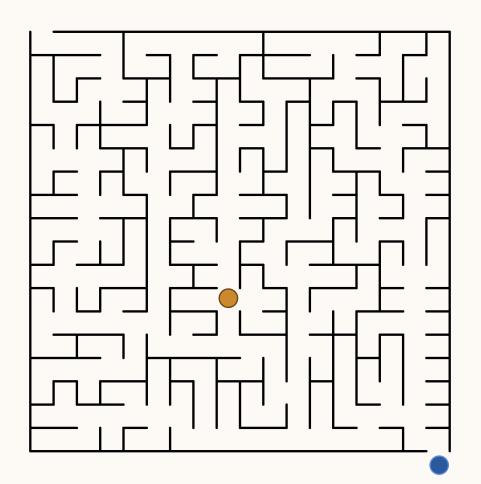
Goal: checkmate the opponent before they checkmate you

Algorithm: Iterative Deepening DFS, because we're optimizing for fewest moves in a large search space (i.e. memory efficiency matters), and we don't know the optimal solution depth. Note that we will have to make some changes to account for the opponent, which we'll talk about when we get to Adversarial Search!



EFFECTIVENESS OF UNINFORMED SEARCH

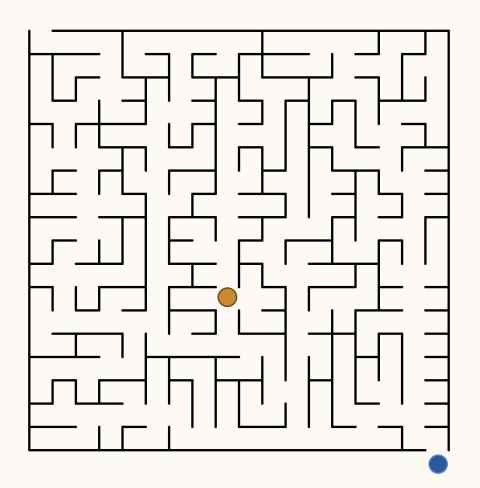
- Where do our search algorithms fit on the acting/thinking humanly/rationally spectrum?
- Consider the maze on the right
 - Find the shortest path from the orange start point to the blue goal point
 - You can move left, right, up, or down
 - Which directions would an uninformed search algorithm consider first?
 - Which directions would *you*, a human, consider first?



EFFECTIVENESS OF UNINFORMED SEARCH

Levels of AI performance:

- Sub-human: can solve a problem, but not as well as a person
- Human: comparable performance to a human (goal of the Turing test?)
- Super-human: outperforms people on the same problems



EFFECTIVENESS OF UNINFORMED SEARCH

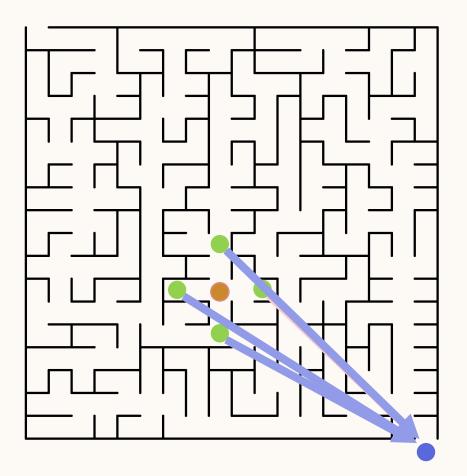
Humans can search some of these toy problems very efficiently, so can we also develop algorithms to solve the same search problems more efficiently?

Informed Search: consider additional information to guide our search to be more efficient; move beyond brute-force techniques

INFORMED SEARCH

INFORMED SEARCH

- down/right states are *closer* to the goal than the up/left states
- Define a function to represent approximate distance to the goal state: a heuristic function h(n)
 - Example: Euclidean (straight-line) distance
- Use h(n) to select nodes for expansion



BEST-FIRST 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

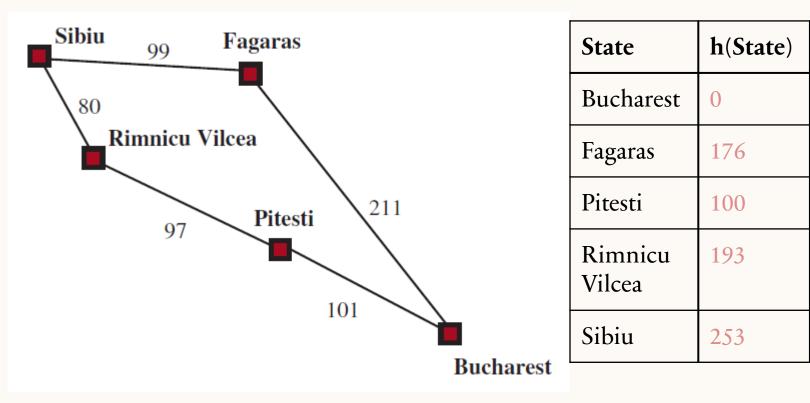
if the node contains a goal state **then return** the corresponding solution expand the chosen node, adding the resulting nodes to the frontier

Store frontier as a priority queue, prioritized by heuristic h(n)

• Heuristic approximates the cost-to-go to reach the goal

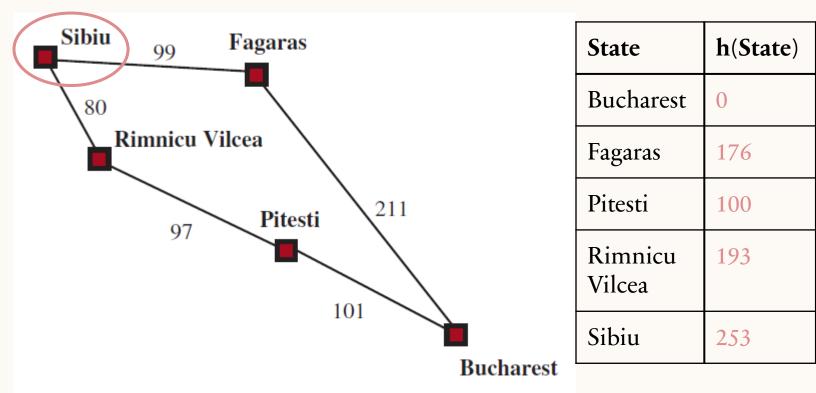
Problem: find the shortest (in terms of distance) path from Sibiu to Bucharest Frontier: [Sibiu (253)]

Current node: *none*



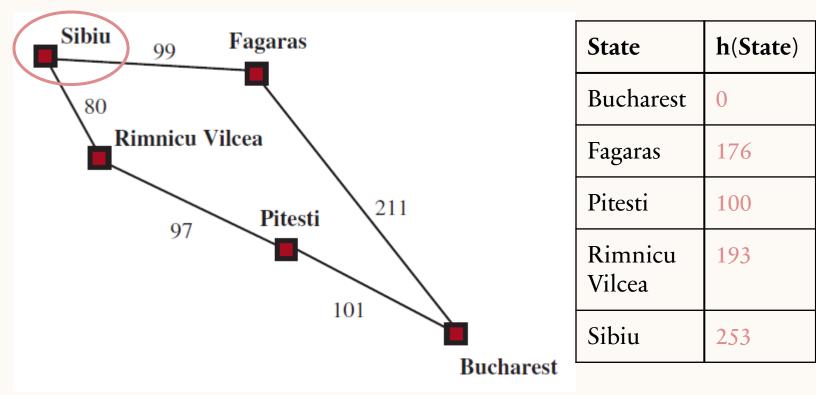
Problem: find the shortest (in terms of distance) path from Sibiu to Bucharest Frontier:

Current node: Sibiu (253)



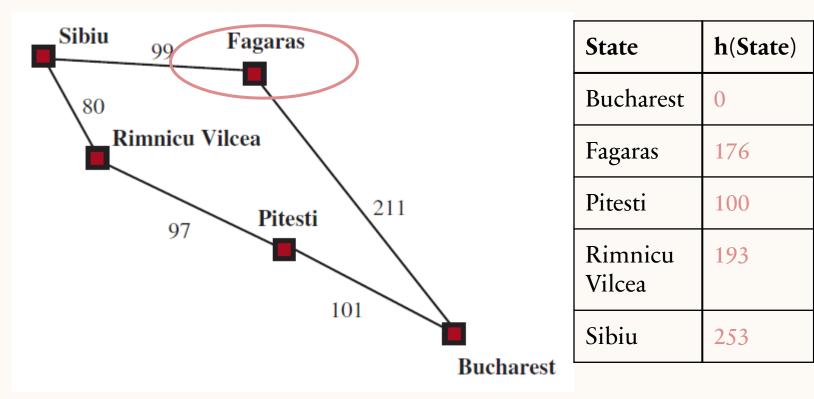
Problem: find the shortest (in terms of distance) path from Sibiu to Bucharest Frontier: [F. (176), R.V. (193)]

Current node: Sibiu (253)



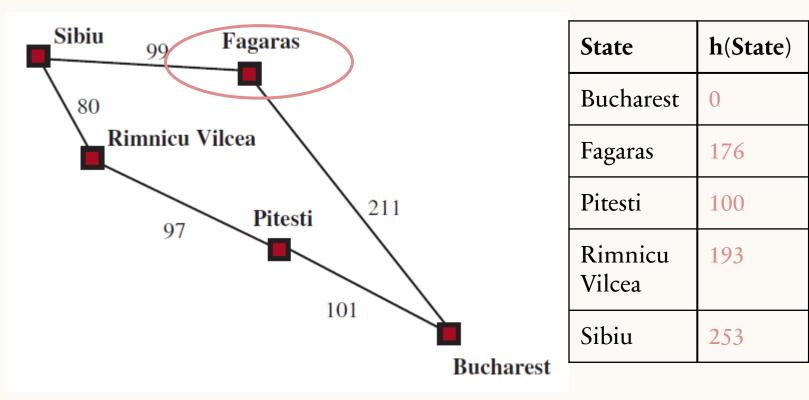
Problem: find the shortest (in terms of distance) path from Sibiu to Bucharest Frontier: [R.V. (193)]

Current node: F. (176)



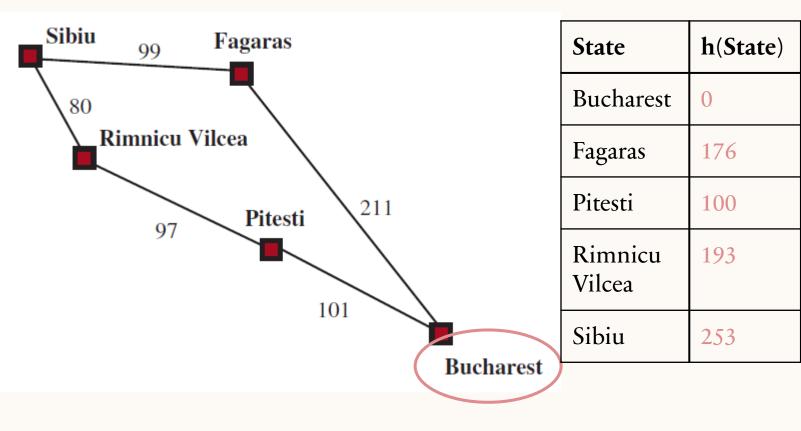
Problem: find the shortest (in terms of distance) path from Sibiu to Bucharest Frontier: [B. (0), R.V. (193)]

Current node: F. (176)



Problem: find the shortest (in terms of distance) path from Sibiu to Bucharest Frontier: [R.V. (193)]

Current node: B. (0) Return: [Sibiu, Fagaras, Bucharest]



BEST-FIRST SEARCH IN PRACTICE

- Experiment URL (try it for yourself!): <u>https://qiao.github.io/PathFinding.js/visual/</u> Experiment summary:
- Best-First Search is a greedy algorithm
 - Every step moves "closer" towards the goal
 - Can miss the bigger picture
- Can be *very* fast with good heuristic choice compared to uninformed search
- May not be optimal depending on the environment

FOR NEXT CLASS

- Read Chapter 3.5-3.7
- If you have Module 1/Search for your paper presentation, start looking for a paper and send it to me & Aydin
 - Presentation on Thursday Sept 21
- Work on the uninformed part of HW 1