

Game AI

Where is the AI

- Route planning / Search
- Movement
- Group behavior
- Decision making

General Search Algorithm Design

- Keep a pair of set of states:
 - One, the set of states to explore, called the **open set** or the *frontier*.
 - The second, the set of states you have seen, called the **closed set**.
- Initially just put the start node in the open set
- While the open set is not empty
 - Take a node from the open set. Add it to the closed set.
 - If the node is a goal node you're done!
 - if next is in closed then continue.
 - Necessary? Not if we check if an item is in Open before adding
 - Otherwise **expand** the node, **generating** all of the node's **successors** and add the ones we have "not seen" to the open list
- If finish the loop because the open set becomes empty, then failure.
- What if one of the successors was a goal node?
- Could we have just declared success right away?
- What order should we remove items to the open set?
- Is our algorithm **complete**? Is it **optimal**?

Basic Search Algorithms

- Depth First (DFS)
 - Organize open set as a stack (LIFO)
- Breadth First (BFS)
 - Organize open set as a queue (FIFO)
- Data structure for closed set?
 - What are the operations? Add and check membership.
- Advantages?
- Complete?
- Optimal?

Breadth First Search (modified)

- Breadth first search can be made a little more efficient:
 - If start is a goal then success!
 - Add start to open
 - While open set is not empty
 - Take the next node off of open.
 - If in closed set continue
 - Generate its successors and for each successor
 - If it is a goal state then done.
 - Otherwise if not seen add to open.
- Where is the gain? Seeing if a state is the goal when we *generate* it.

Saving space

- BFS has a large frontier / open set.
 - It grows exponentially.
- Can we reduce it?
- Bi-directional
- Bounded DFS
- Iterative DFS

Bidirectional

- Apply a breadth first search from both start and from goal.
- When their “frontiers” intersect we have a solution.
- Benefit?
- Space reduced from $O(bd)$ to $O(bd/2)$

Depth Limited

- If you know the solution can not be any deeper than depth k
- Then use DFS and cut off your search at depth k .
- May greatly speed up DFS.
- Ensures completeness.
- Optimal?

Iterative Deepening

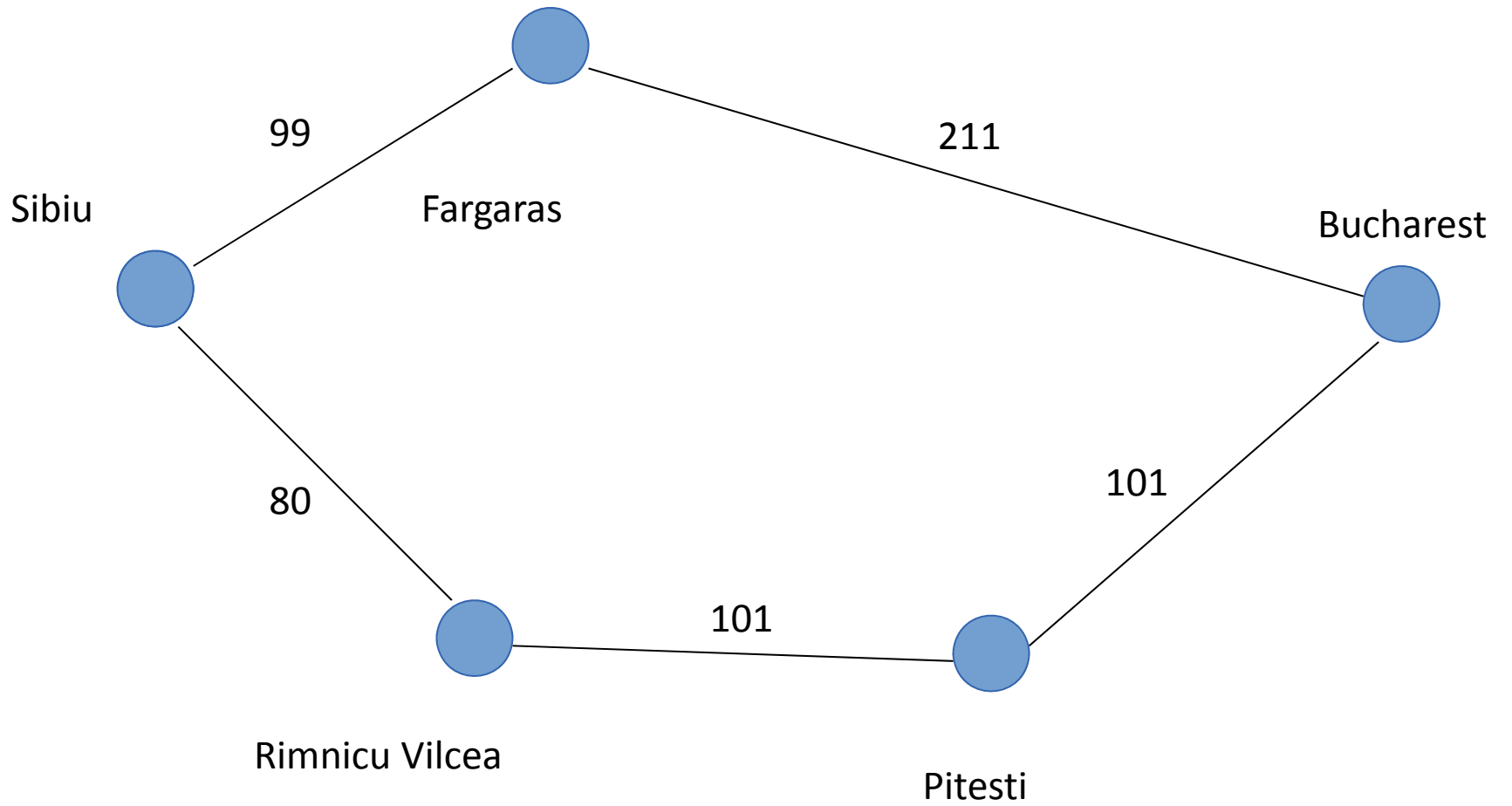
- for (int level = 0; found?; ++level)
 depthLimited(level)
- Huh? We will be searching the beginning levels over and over!
All our efforts will be thrown out!
- Yes, compared to BFS we are trading time for space.
- But it is optimal!

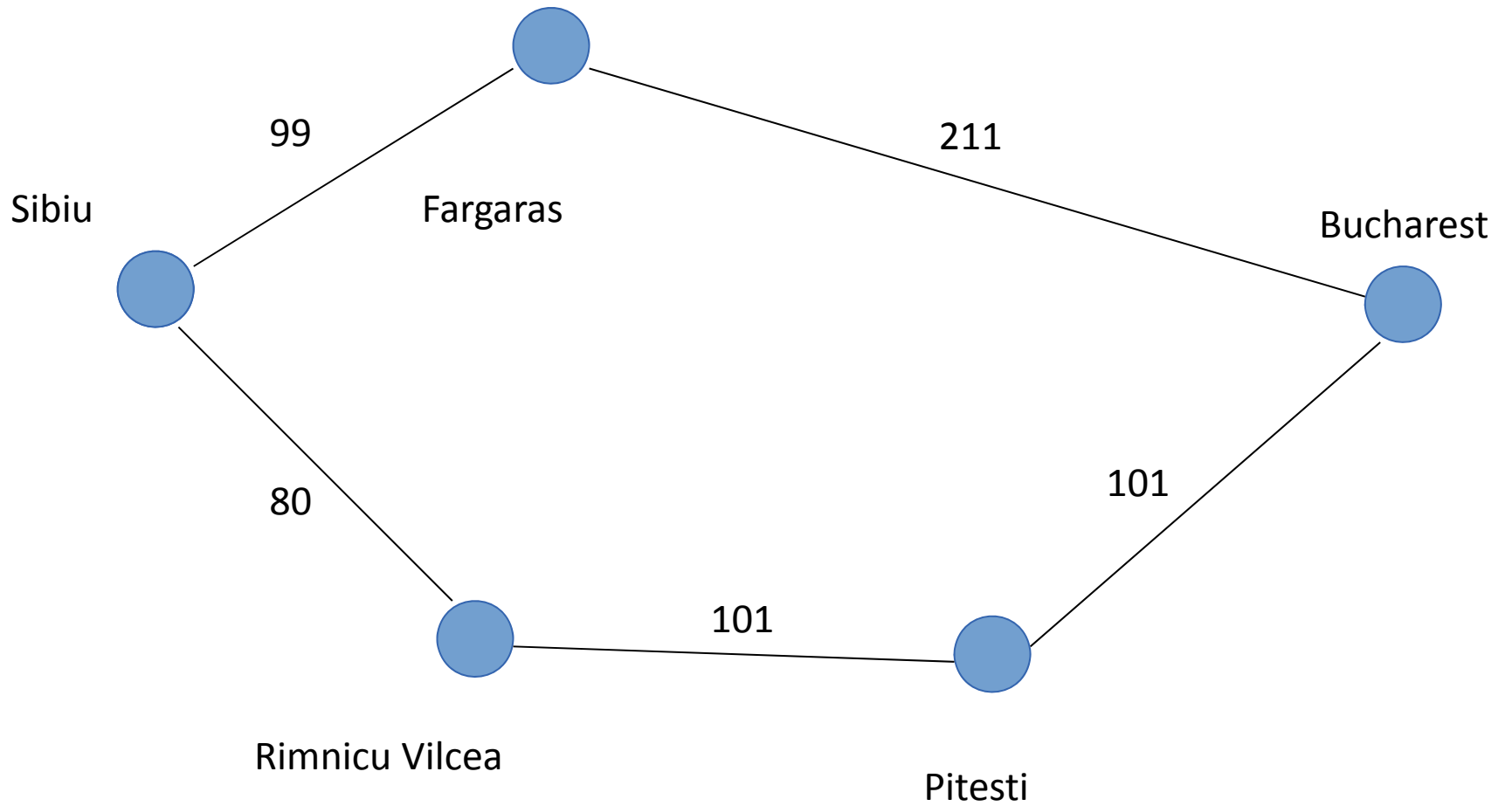
Uniform Cost

- What if the effort to go from one state to another is not always the same?
- E.g. Traveling on a diagonal in a grid might cost 1.4 times the cost of left/right/up/down.
- Or our “steps” might involve plane trips of different distances.
- “Uniform cost” means we want to expand our search so that we explore nodes *uniformly* in how much it costs to get to them.
- Same as Dijkstra’s shortest path, but that is to all nodes, not just to a goal.
- Critical point in the algorithm:
 - a state on the open set may change!
 - Cannot use optimization that we used in BFS

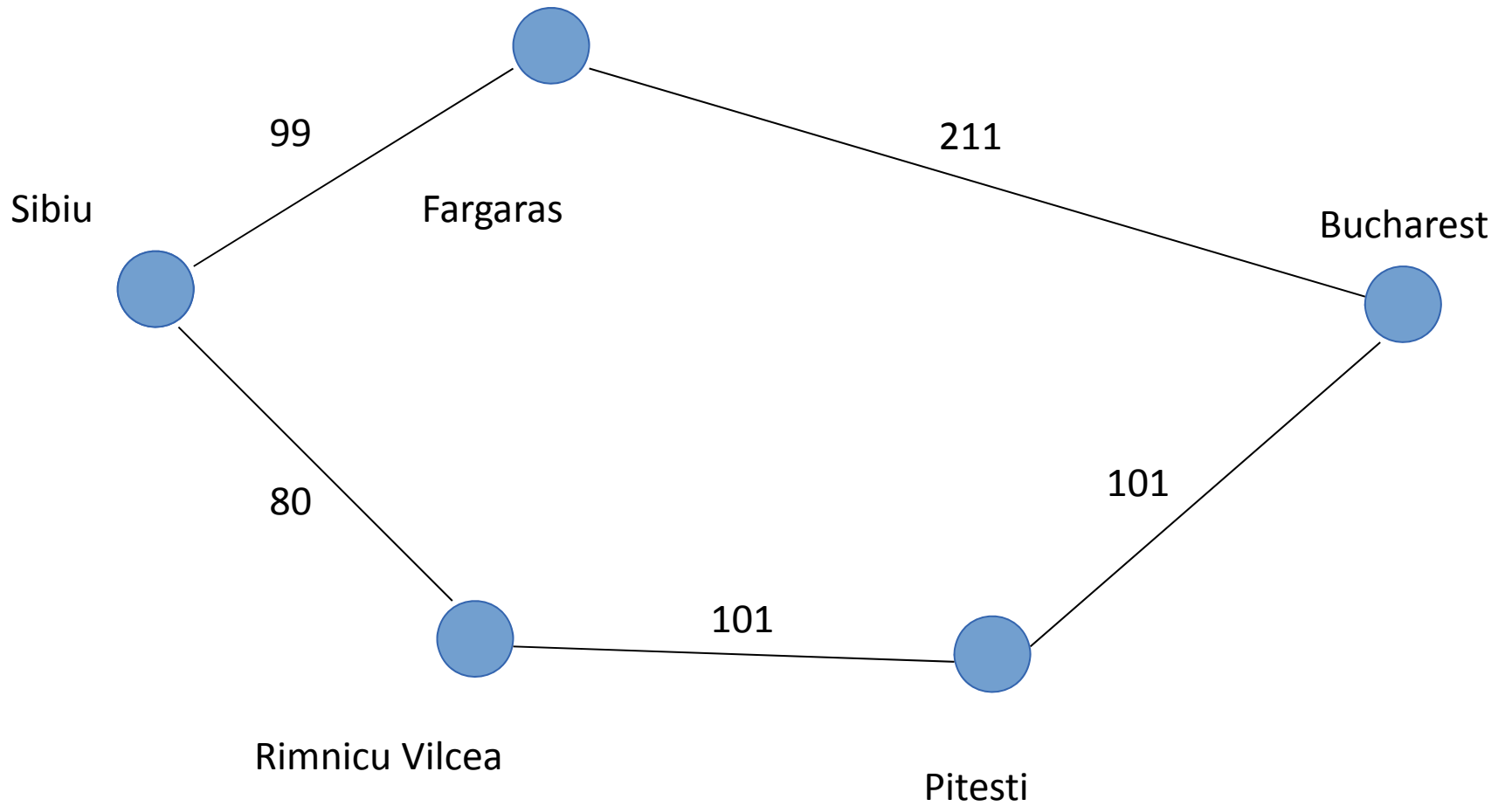
From Sibiu to Bucharest

| | | |
|----------------|----------------|-----|
| | | |
| Sibiu | Fagaras | 99 |
| Sibiu | Rimnicu Vilcea | 80 |
| Fargaras | Bucharest | 211 |
| Rimnicu Vilcea | Pitesti | 101 |
| Pitesti | Bucharest | 101 |

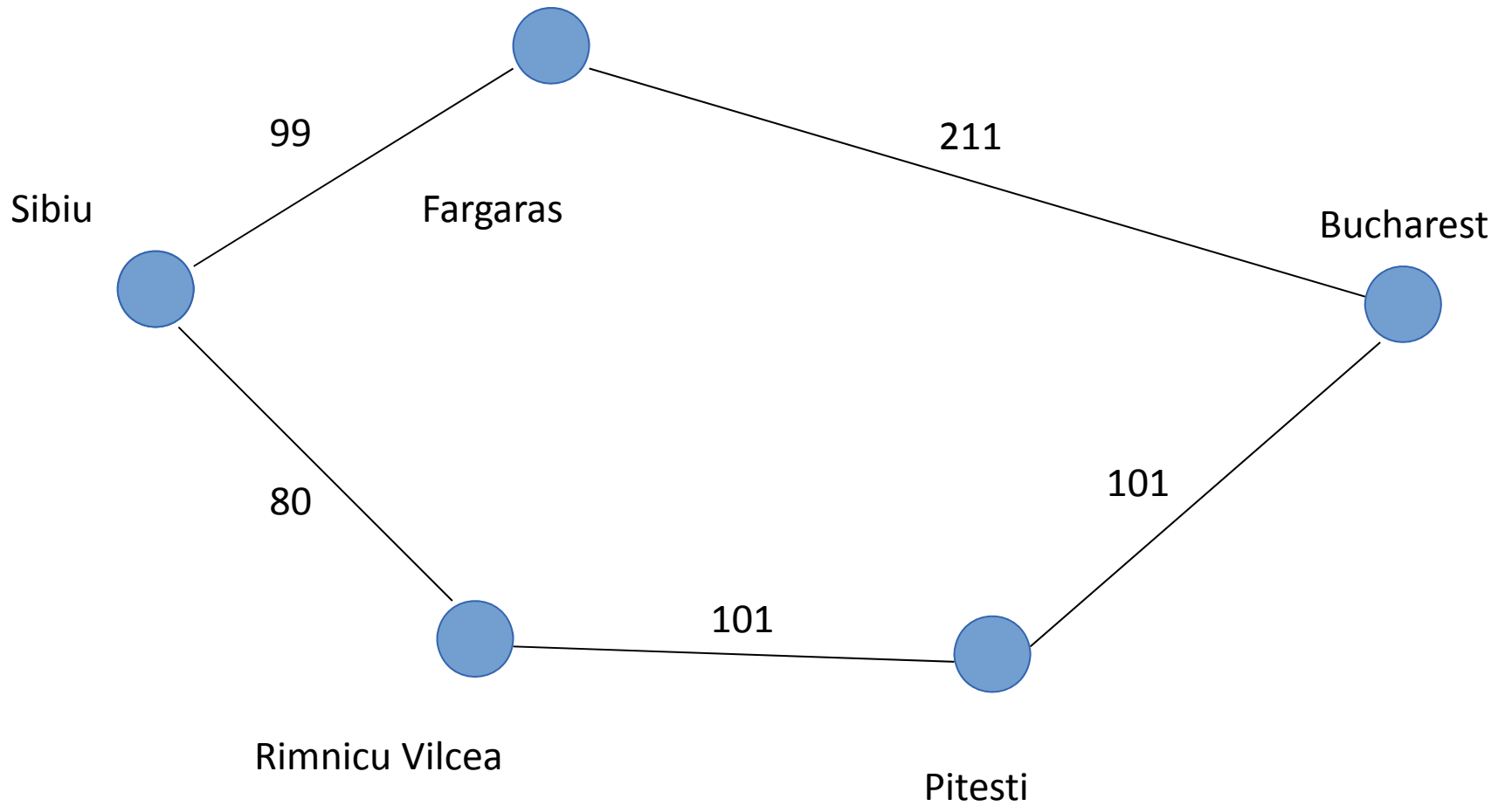




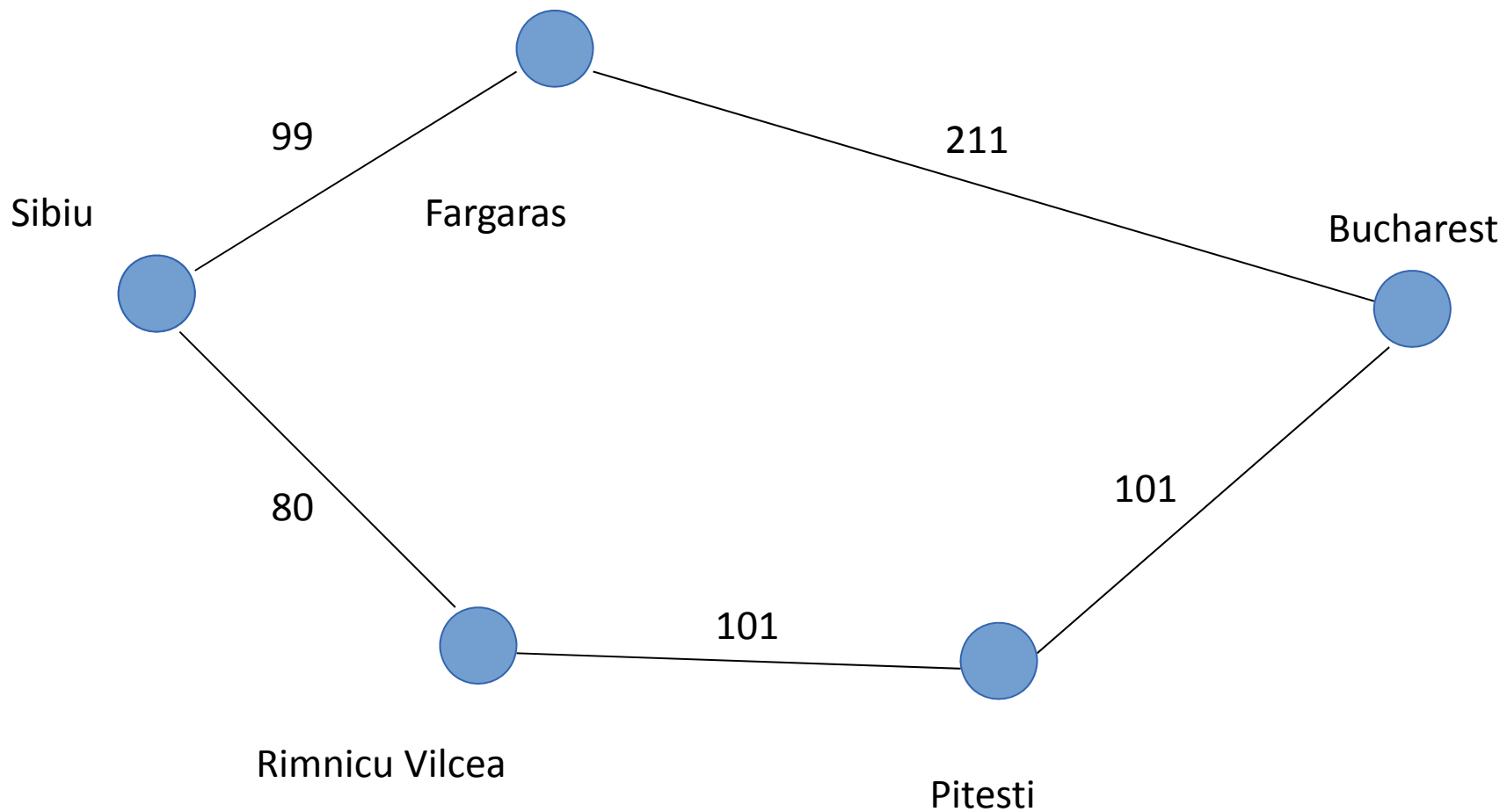
Open: [S(0)]



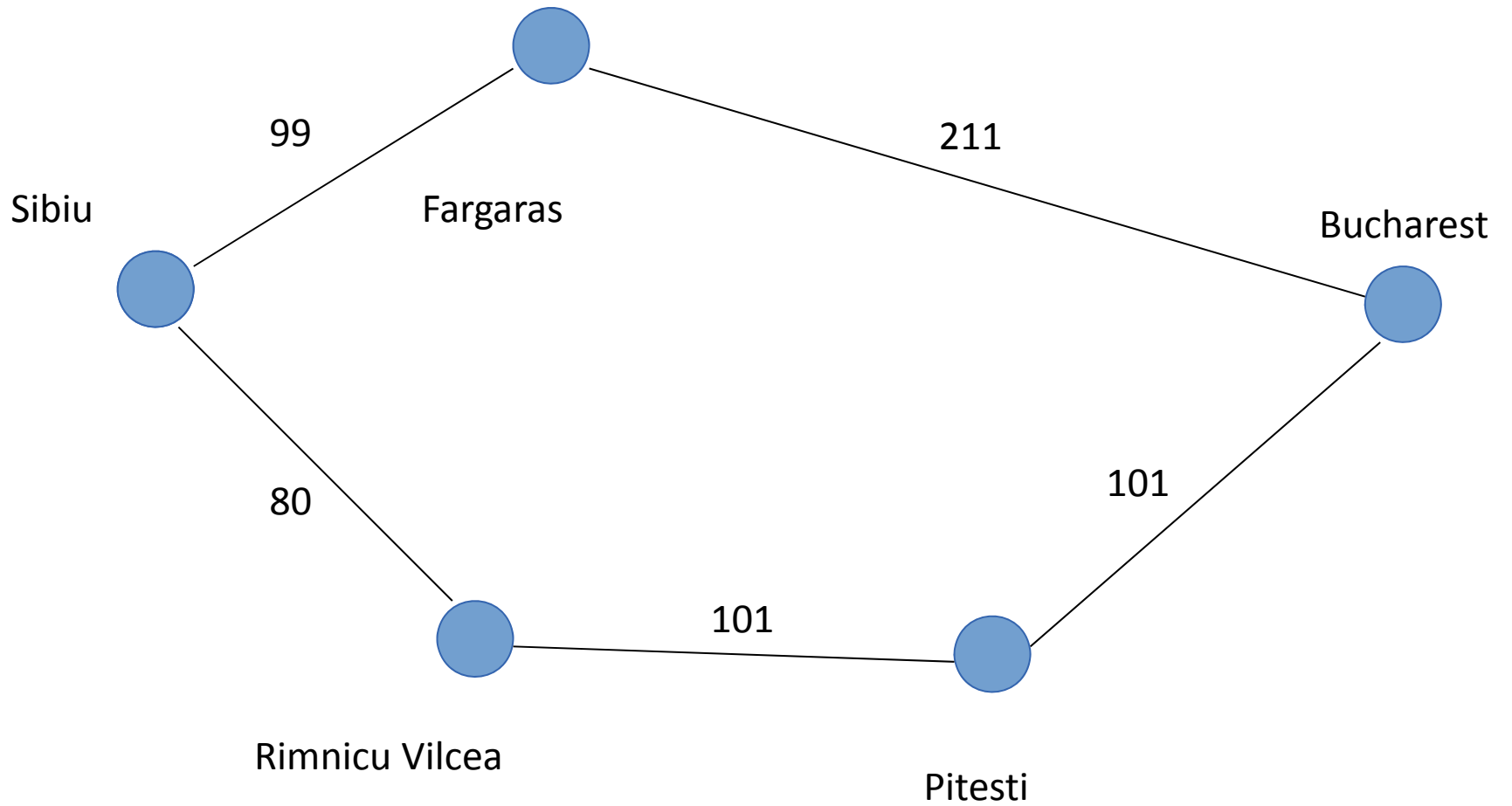
Open: [R(80), F(99)]



Open: [F(99), P(181)]



Open: [P(181), B(310)]



Open: [B(282)]

Uninformed Search

- Uninformed?
- Means we don't know how "far" it is to the goal.
- Depth First
 - Uses a stack to represent open list.
- Breadth First
 - Uses a queue to represent open.
- Uniform Cost / Dijkstra (for a single goal)
 - Add nodes based on current cost of reaching the node.
 - i.e. use a priority queue. Similar to breadth first, but does not assume every step has the same cost.
- Variations
 - Depth Limited
 - Iterative Deepening
 - Bidirectional

Informed

- Suppose we have *some* idea how “far” the goal is away from each node?
- The idea is known as a **heuristic**.
- It is not assumed to be accurate.
- We could always expand a node that promises to get us closest to the goal.
- That’s called being **greedy**.
- Or we could also take into account the actual cost to get us as far as we have come.
- That’s known as **A* algorithm**

A*

- Data Structures:

- start node

- target node

- successor function

- For each node:

- Path taken (e.g. pointer to prior)

- Cost from start: $g(n)$

- Estimated cost to goal: $h(n)$,

- open list (**ordered by cost function $f(n) = g(n) + h(n)$**)

- closed set

A*

- Add start node to open list
- While open list is not empty
 - remove highest priority node (lowest **estimated** cost)
 - if node is goal, then success: return its solution path.
 - Else
 - place node on closed list
 - generate successors. For each successor:
 - Compute cost: **$f(n) = g(n) + h(n)$**
 - If successor is on open and new $g(n)$ is better than old,
 - update entry on open
 - Else **if successor is on closed and new $g(n)$ is better than old,**
 - **remove from closed and add to open with new $f(n)$**
 - Else if not previously seen
 - add to open.
- If the queue becomes empty then there was no solution.

Admissibility

- A heuristic is called **admissible** if it is guaranteed to be an underestimate of the actual distance for all cases
- An admissible heuristic will result in A* being optimal.

A* in Problem Solving

- 15-Puzzle

- Heuristics

- 1) How many tiles are out of position?

- 2) What is the total *distance* that tiles are out of place?

- Path-finding

- Heuristics

- 1) Crow flies distance

- 2) Manhattan distance.

Triangle Inequality

- **Consistent** (aka **monotonic**)
- A heuristic function h is consistent if
 - Given nodes n_1 and n_2
 - And their heuristic costs $h(n_1)$ and $h(n_2)$
 - Together with the actual cost to go from n_1 to n_2 , $c(n_1, n_2)$
 - Then: $h(n_1) \leq h(n_2) + c(n_1, n_2)$
- If a heuristic function is consistent then when a node is placed on the closed set, it never has to move back to the open set.
- Admissible functions are *almost always* consistent.
- But games like to use inadmissible functions.

Data Structures for Open Set

| | Insert | Membership | Get/Remove Best | Adjust |
|----------------------|--------------|----------------------------|---------------------|------------------------------------------|
| Unsorted Array/List | $O(1)$ | $O(F)$ | $O(F)$ | $O(F)$ |
| Sorted Array | $O(F)$ | Binary search: $O(\log F)$ | Keep at end: $O(1)$ | Find: $O(\log F)$ Change: $O(F)$ |
| Sorted Linked List | Find: $O(F)$ | $O(F)$ | $O(1)$ | $O(F)$ for find. $O(1)$ to adjust |
| Indexed Array | $O(1)$ | $O(1)$ | $O(N)$ | $O(1)$ |
| Hash Table | $O(1)$ | $O(1)$ | $O(N)$ | $O(1)$ |
| Heap | $O(\log F)$ | $O(F)$ | $O(\log F)$ | $O(F)$ for find $O(\log F)$ to adjust |
| Heap + Indexed Array | $O(\log F)$ | $O(1)$ | $O(\log F)$ | $O(F)$ |