

# CSC-411

# Artificial

# Intelligence

Search Agents





# State Space vs Search Space

- **State space:** a physical configuration
- **Search space:** an abstract configuration represented by a search tree or graph of possible solutions
- **Search tree:** models the sequence of actions
  - **Root:** initial state
  - **Branches:** actions
  - **Nodes:** results from actions. A node has: parent, children, depth, path cost, associated state in the state space.
- **Expand:** A function that given a node, creates all children nodes

# Search Space Regions

- The Search space is divided into three regions:
  - **Explored** (a.k.a. Closed List, Visited Set)
  - **Frontier** (a.k.a. Open List, the Fringe)
  - **Unexplored**
- The essence of search is moving nodes from regions (3) to (2) to (1), and the essence of search strategy is deciding the order of such moves.



# Tree Search

**function** TREE-SEARCH(initialState, goalTest)  
*returns* **SUCCESS** or **FAILURE** :

```
initialize frontier with initialState
```

```
while not frontier.isEmpty():
    state = frontier.remove()
```

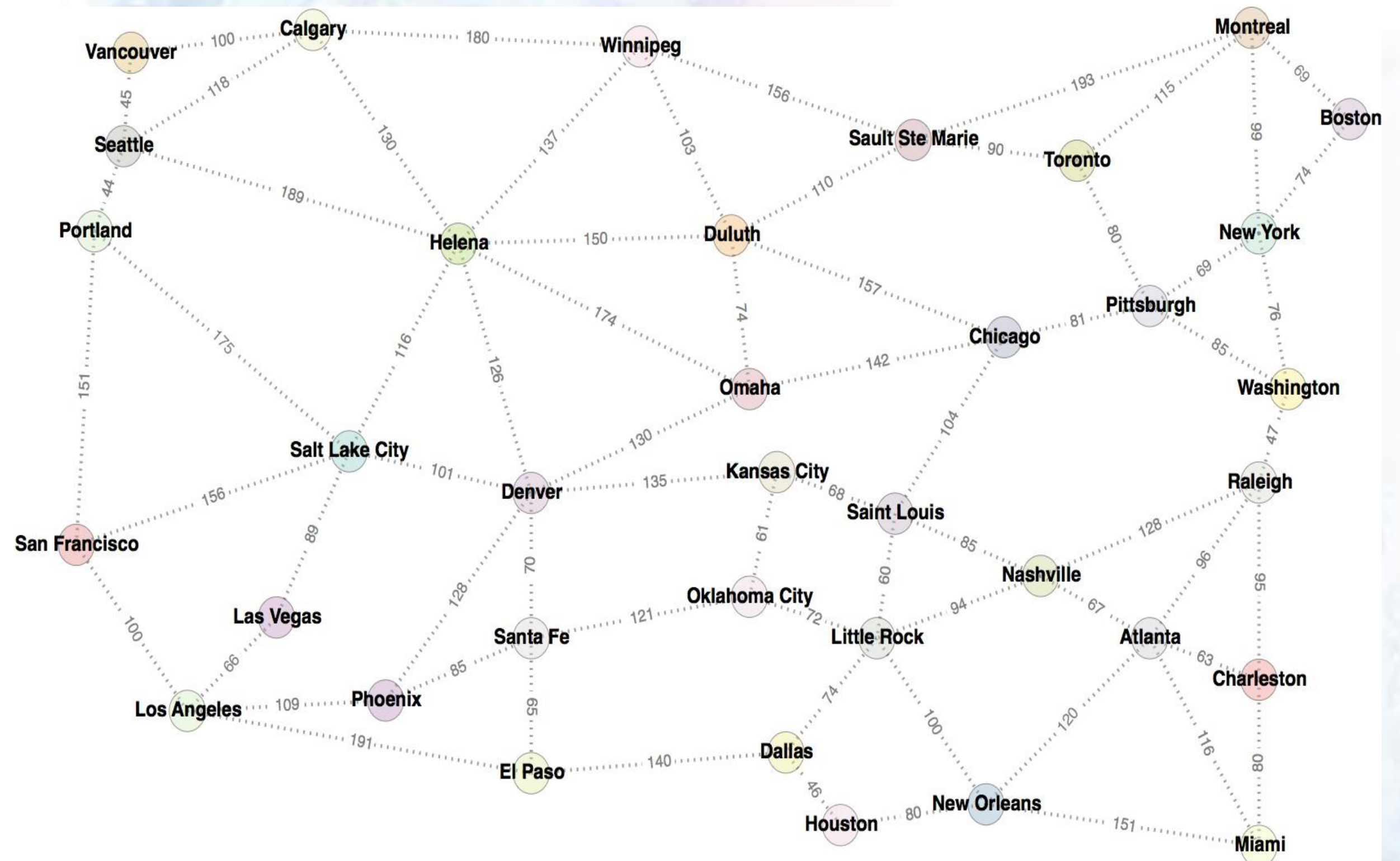
```

if goalTest(state):
    return SUCCESS(state)

```

```
for neighbor in state.neighbors():
    frontier.add(neighbor)
```

```
return FAILURE
```



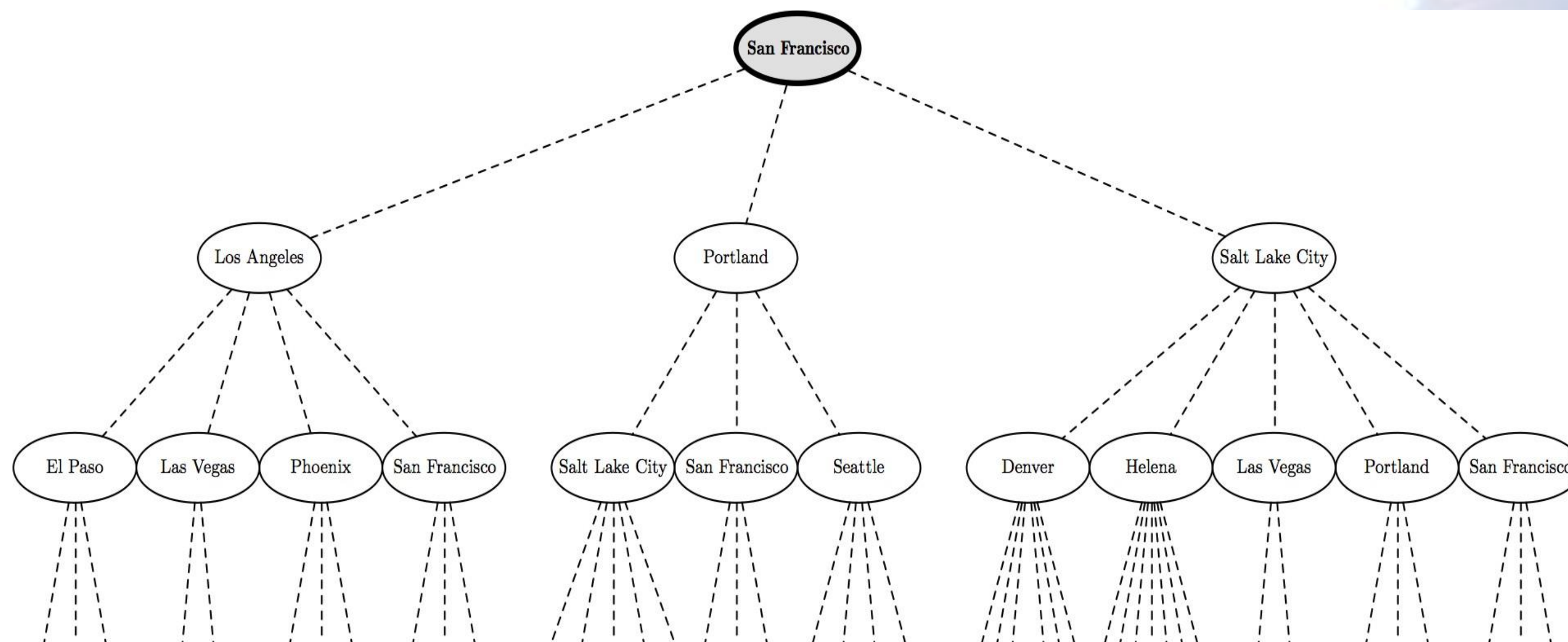
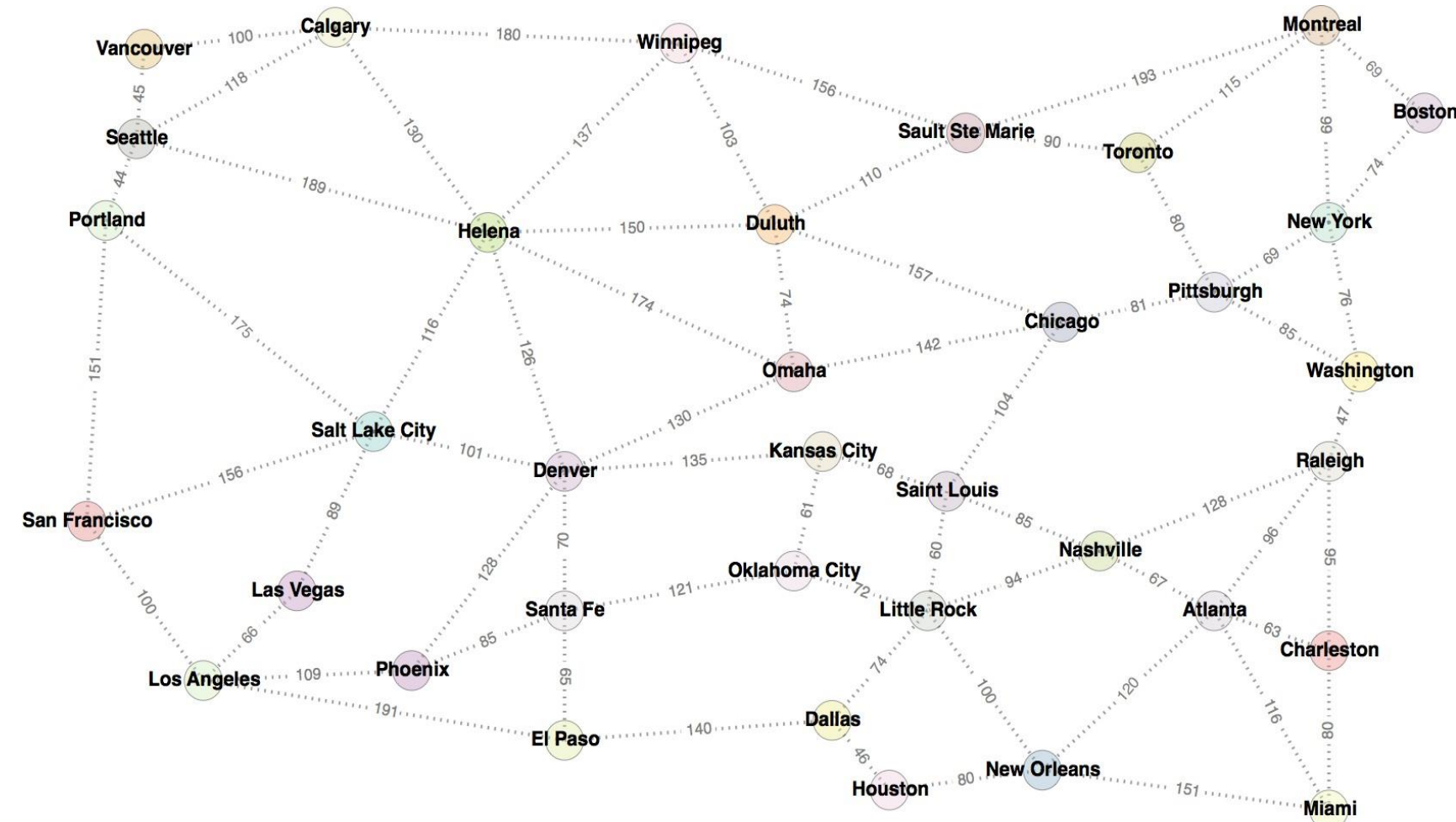


# Tree Search Example

- Let's show the first steps in growing the search tree to find a route from San Francisco to another city.
- In the following we adopt the following color coding: orange nodes are explored, grey nodes are the frontier, white nodes are unexplored, and black nodes are failures.



# Tree Search Example



**function** TREE-SEARCH(initialState, goalTest)  
*returns* **SUCCESS** or **FAILURE** :

**initialize** frontier **with** initialState

**while not** frontier.isEmpty():  
     state = frontier.remove()

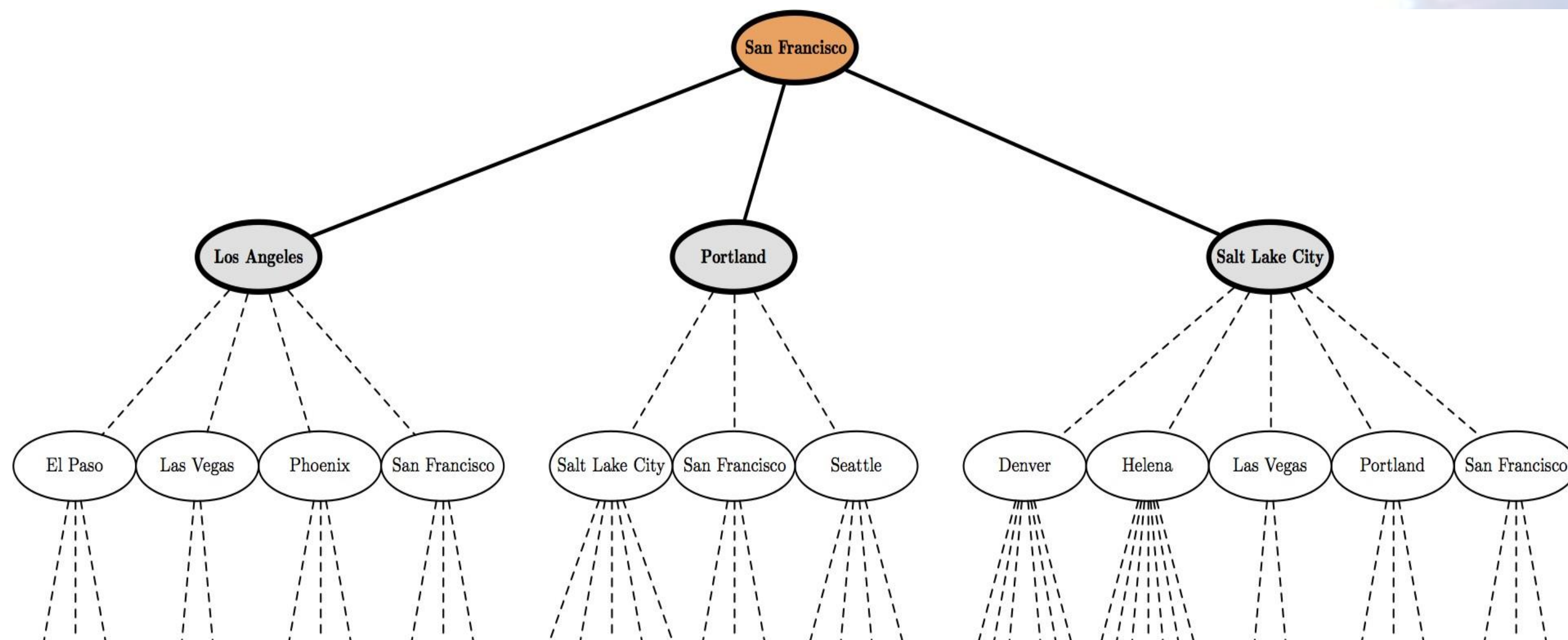
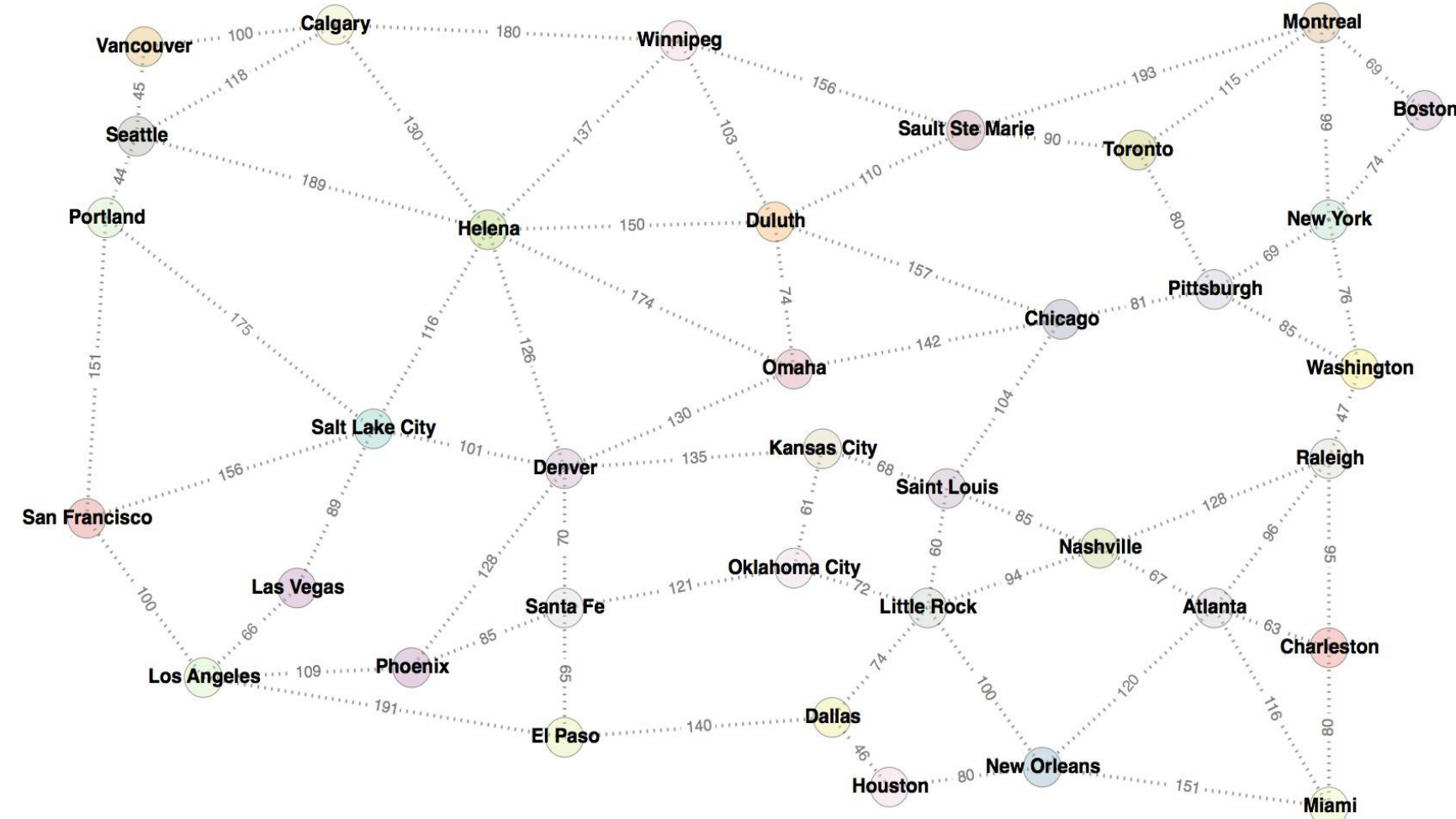
**if** goalTest(state):  
     **return** **SUCCESS**(state)

**for** neighbor **in** state.neighbors():  
     frontier.add(neighbor)

**return** **FAILURE**



# Tree Search Example



**function** TREE-SEARCH(initialState, goalTest)  
*returns* **SUCCESS** or **FAILURE** :

**initialize** frontier **with** initialState

**while not** frontier.isEmpty():  
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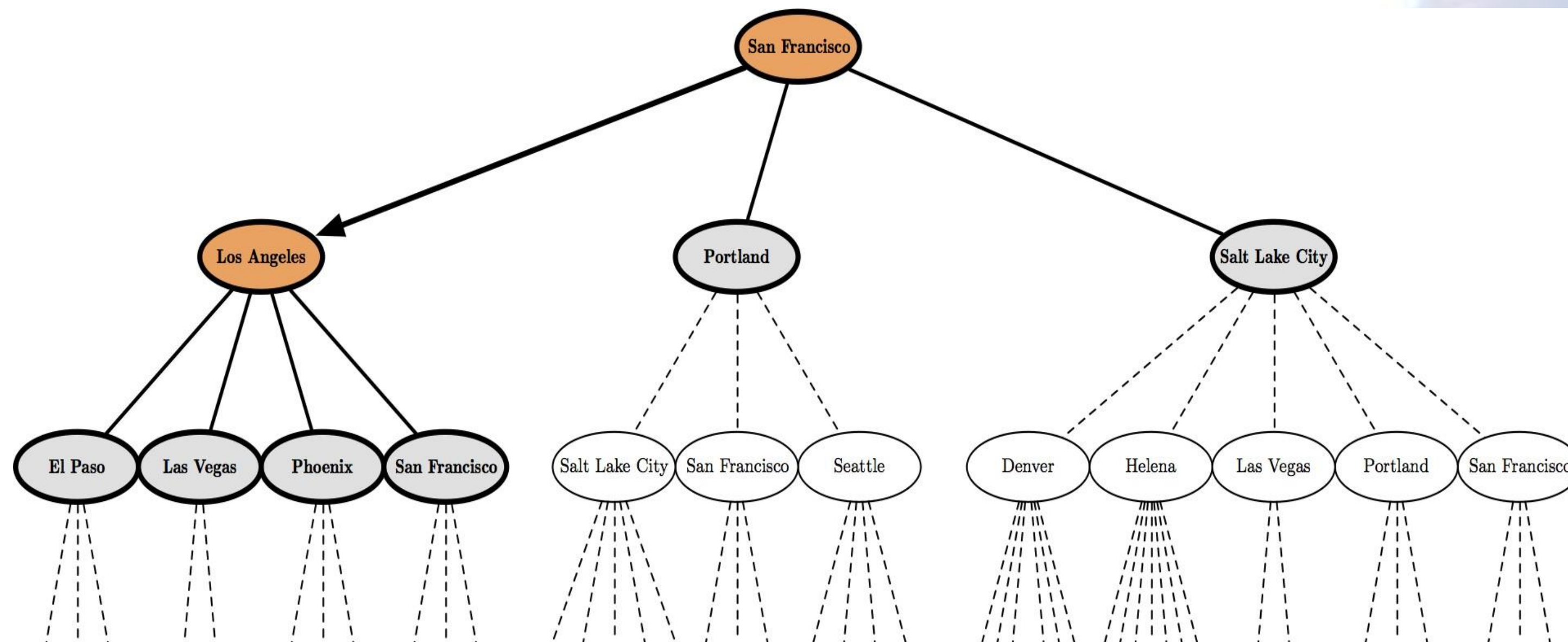
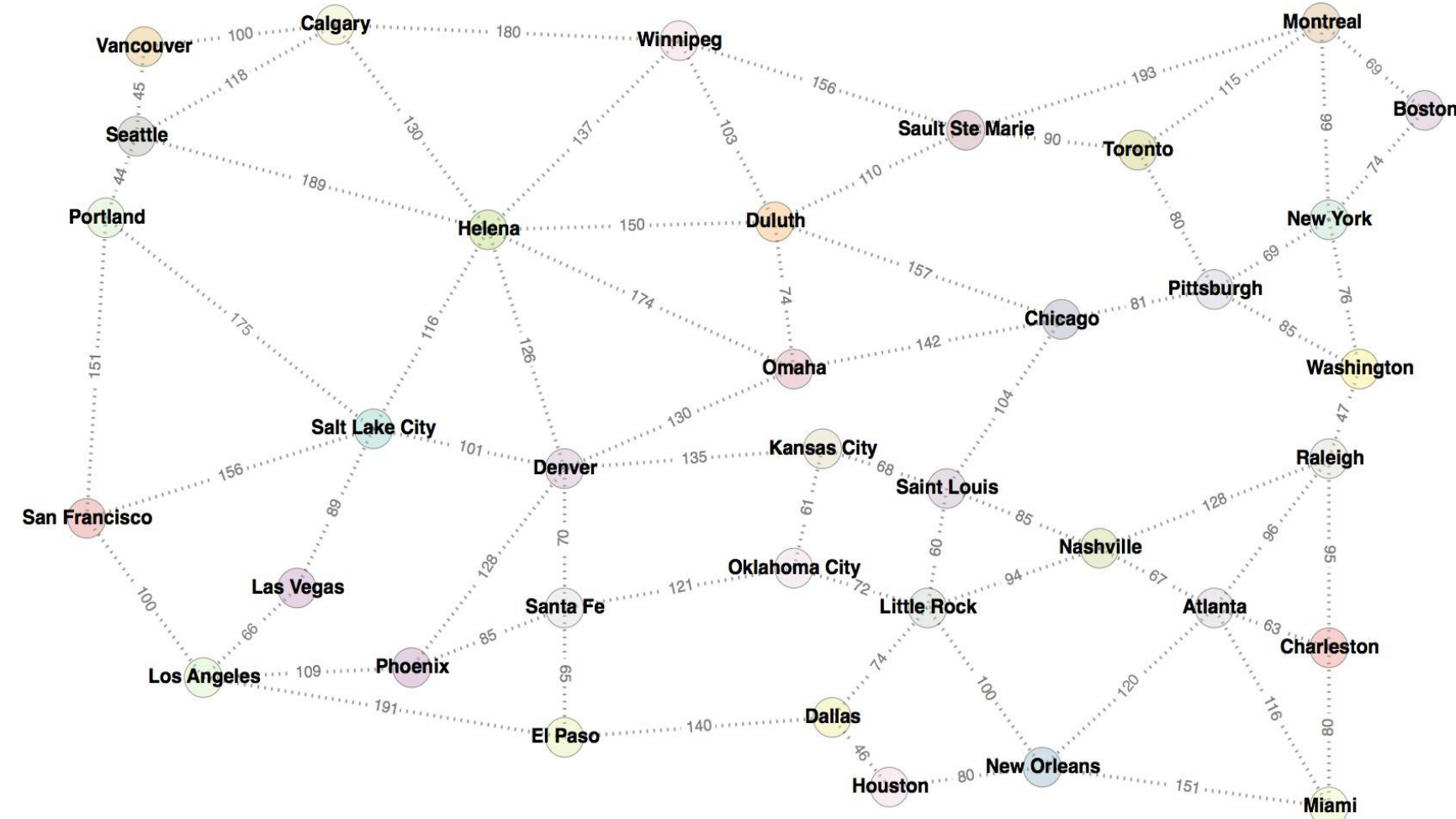
**if** goalTest(state):  
     **return** **SUCCESS**(state)

**for** neighbor **in** state.neighbors():  
     frontier.add(neighbor)

**return** **FAILURE**



# Tree Search Example



**function** TREE-SEARCH(initialState, goalTest)  
*returns* **SUCCESS** or **FAILURE** :

**initialize** frontier **with** initialState

**while not** frontier.isEmpty():  
     state = frontier.remove()

**if** goalTest(state):  
     **return** **SUCCESS**(state)

**for** neighbor **in** state.neighbors():  
     frontier.add(neighbor)

**return** **FAILURE**



# Graph Search

**function** GRAPH-SEARCH(initialState, goalTest)  
*returns* **SUCCESS** or **FAILURE** :

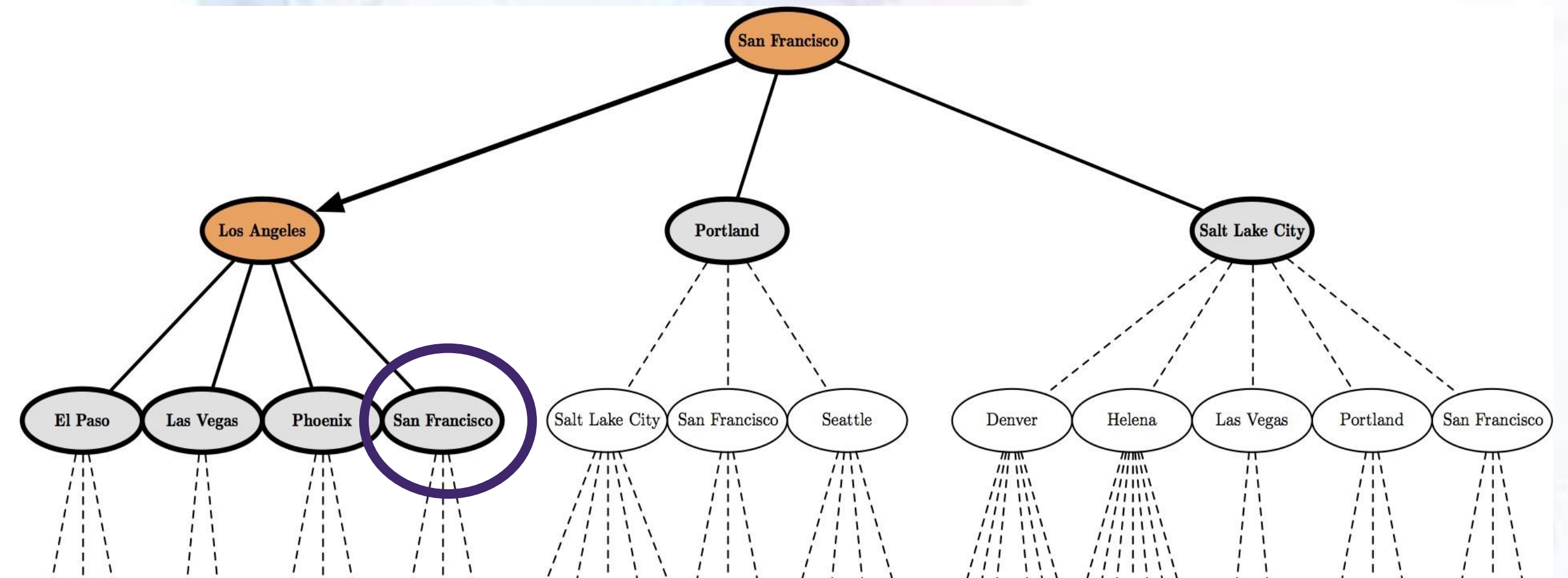
**initialize** frontier **with** initialState  
 explored = Set.new()

**while not** frontier.isEmpty():  
   state = frontier.remove()  
   explored.add(state)

**if** goalTest(state):  
   return **SUCCESS**(state)

**for** neighbor **in** state.neighbors():  
   **if** neighbor **not in** frontier  $\cup$  explored:  
     frontier.add(neighbor)

return **FAILURE**





# Search Strategies

- A strategy is defined by picking the order of node expansion
- Strategies are evaluated along the following dimensions:
  - Completeness
    - Does it always find a solution if one exists?
  - Time complexity
    - Number of nodes generated/expanded
  - Space complexity
    - Maximum number of nodes in memory
  - Optimality
    - Does it always find a least-cost solution?



# Search Strategies

- Time and space complexity are measured in terms of:
  - $b$ : maximum branching factor of the search tree (actions per state).
  - $d$ : depth of the solution
  - $m$ : maximum depth of the state space (may be  $\infty$ ) (also noted sometimes  $D$ ).
- Two kinds of search: **Uninformed** and **Informed**



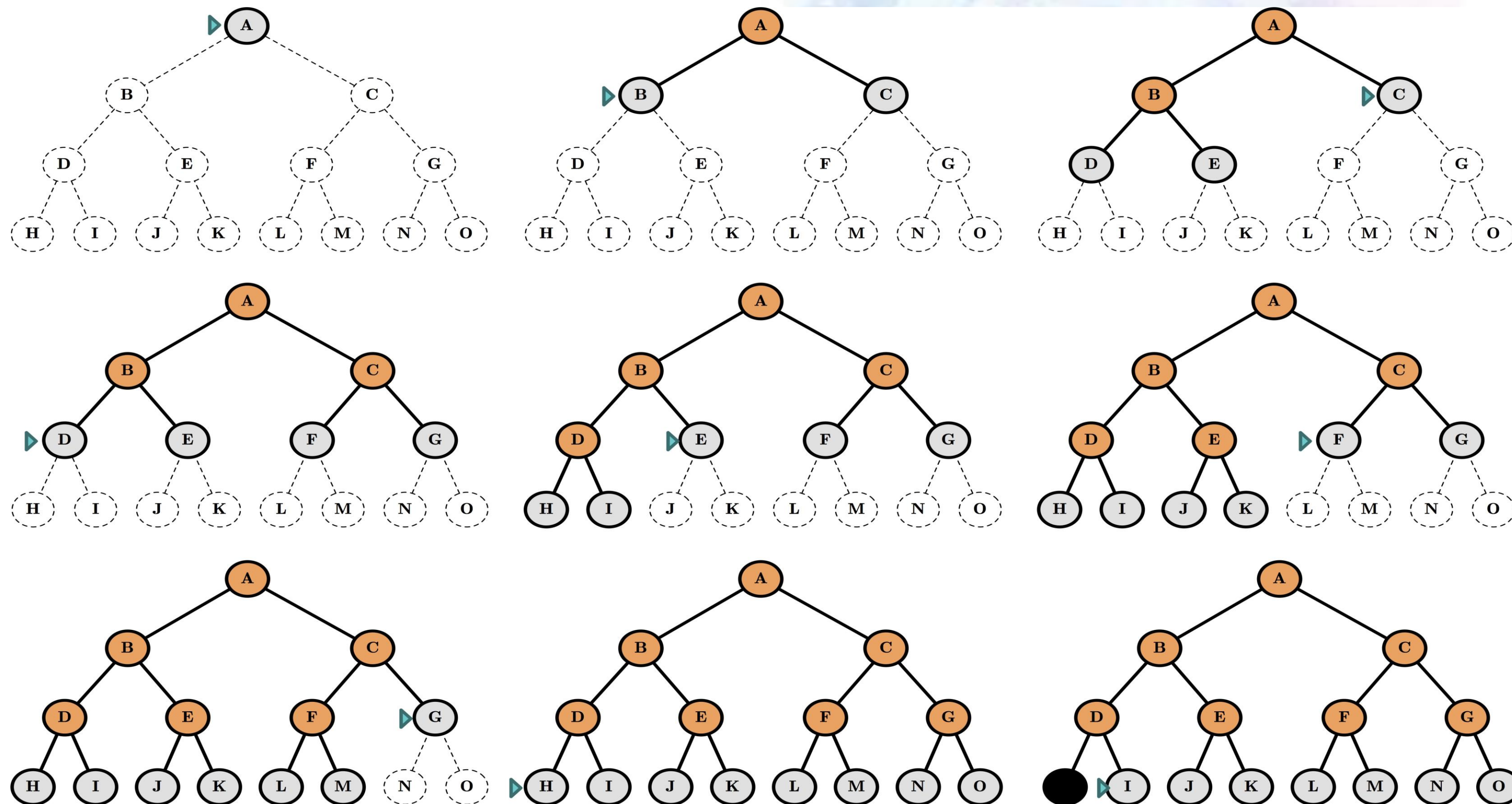
# Uninformed Search

- **Use no domain knowledge!**
- **Strategies:**
  1. **Breadth-first search (BFS):** Expand shallowest node
  2. **Depth-first search (DFS):** Expand deepest node
  3. **Depth-limited search (DLS):** Depth first with depth limit
  4. **Iterative-deepening search (IDS):** DLS with increasing limit
  5. **Uniform-cost search (UCS):** Expand least cost node



# Breadth-First Search (BFS)

- Use no domain knowledge!





# Breadth-First Search Algorithm

```
function BREADTH-FIRST-SEARCH(initialState, goalTest)
```

```
    returns SUCCESS or FAILURE :
```

```
    frontier = Queue.new(initialState)
```

```
    explored = Set.new()
```

```
    while not frontier.isEmpty():
```

```
        state = frontier.dequeue()
```

```
        explored.add(state)
```

```
        if goalTest(state):
```

```
            return SUCCESS(state)
```

```
        for neighbor in state.neighbors():
```

```
            if neighbor not in frontier  $\cup$  explored:
```

```
                frontier.enqueue(neighbor)
```

```
    return FAILURE
```





# BFS Criteria

- **Complete:** Yes (if  $b$  is finite)
- **Time:**  $1 + b + b^2 + b^3 + \dots + b^d = O(b^d)$
- **Space:**  $O(b^d)$
- **Note:** If the goal test is applied at expansion rather than generation then  $O(b^{d+1})$
- **Optimal:** Yes (if cost = 1 per step)
- **Implementation:** Fringe - FIFO (Queue)



# BFS Criteria

- How bad is BFS?

Depth	Nodes	Time	Memory
2	110	.11 milliseconds	107 kilobytes
4	11,110	11 milliseconds	10.6 megabytes
6	$10^6$	1.1 seconds	1 gigabyte
8	$10^8$	2 minutes	103 gigabytes
10	$10^{10}$	3 hours	10 terabytes
12	$10^{12}$	13 days	1 petabyte
14	$10^{14}$	3.5 years	99 petabytes
16	$10^{16}$	350 years	10 exabytes

- Time and Memory requirements for breadth-first search for a branching factor  $b=10$ ; 1 million nodes per second; 1,000 bytes per node.
- **Memory requirement + exponential time complexity are the biggest handicaps of BFS!**