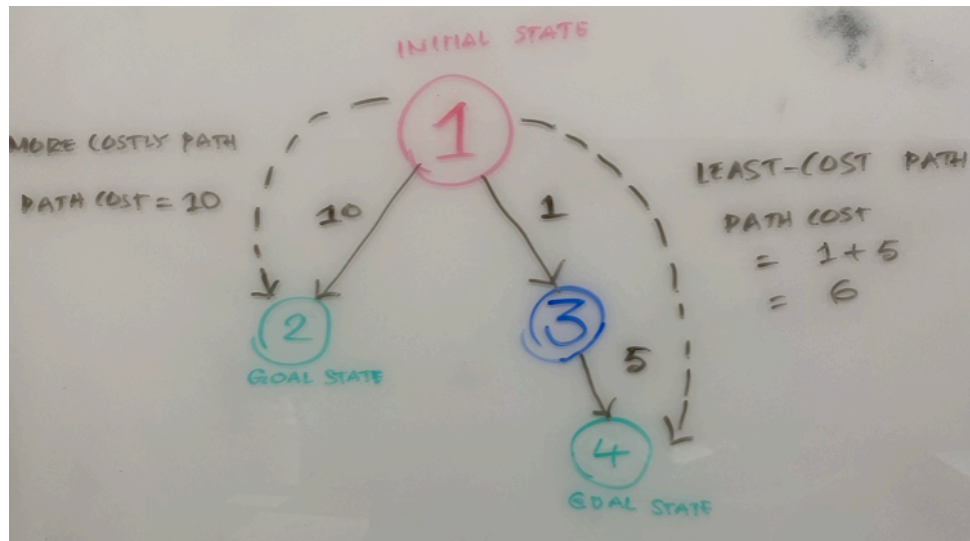


# CS143: Artificial Intelligence



Analysis of Search Algorithms (DFS, IDS)  
Uniform Cost Search (UCS)



# Review: Analysis of BFS

# Properties of Breadth-First Search (BFS)

- **Completeness:**
  - a search algorithm is **complete** if it finds a solution whenever one exists
- **Optimality:**
  - a search algorithm is **optimal** if it returns a minimum-cost path if a solution exists
- **Time complexity:**
  - time required by the algorithm
- **Space complexity:**
  - memory required by the algorithm

# Properties of Breadth-First Search (BFS)

- Is BFS Complete?
- Is BFS Optimal?
- What is the running time complexity of BFS?
- What are the memory requirement of BFS?

# Properties of Breadth-First Search (BFS)

**b**: branching factor

**d**: depth of shallowest goal node

- Is BFS complete?
  - BFS is complete for a finite branching factor **b**
- Is BFS optimal?
  - BFS is optimal if step cost is 1
  - BFS will find the goal node at the shallowest depth before reaching goal nodes at a deeper level

# Properties of Breadth-First Search (BFS)

**b**: branching factor

**d**: depth of shallowest goal node

- Time complexity of BFS?

Number of nodes generated (in the worst-case):

$$1 + b + b^2 + \dots + b^d + b*(b^d - 1) = O(b^{d+1})$$

Time complexity is  $O(b^{d+1})$

- Let's check — how did we arrive at above formula.

# Properties of Breadth-First Search (BFS)

**b**: branching factor

**d**: depth of shallowest goal node

- Memory complexity of BFS?

Number of nodes kept in the frontier

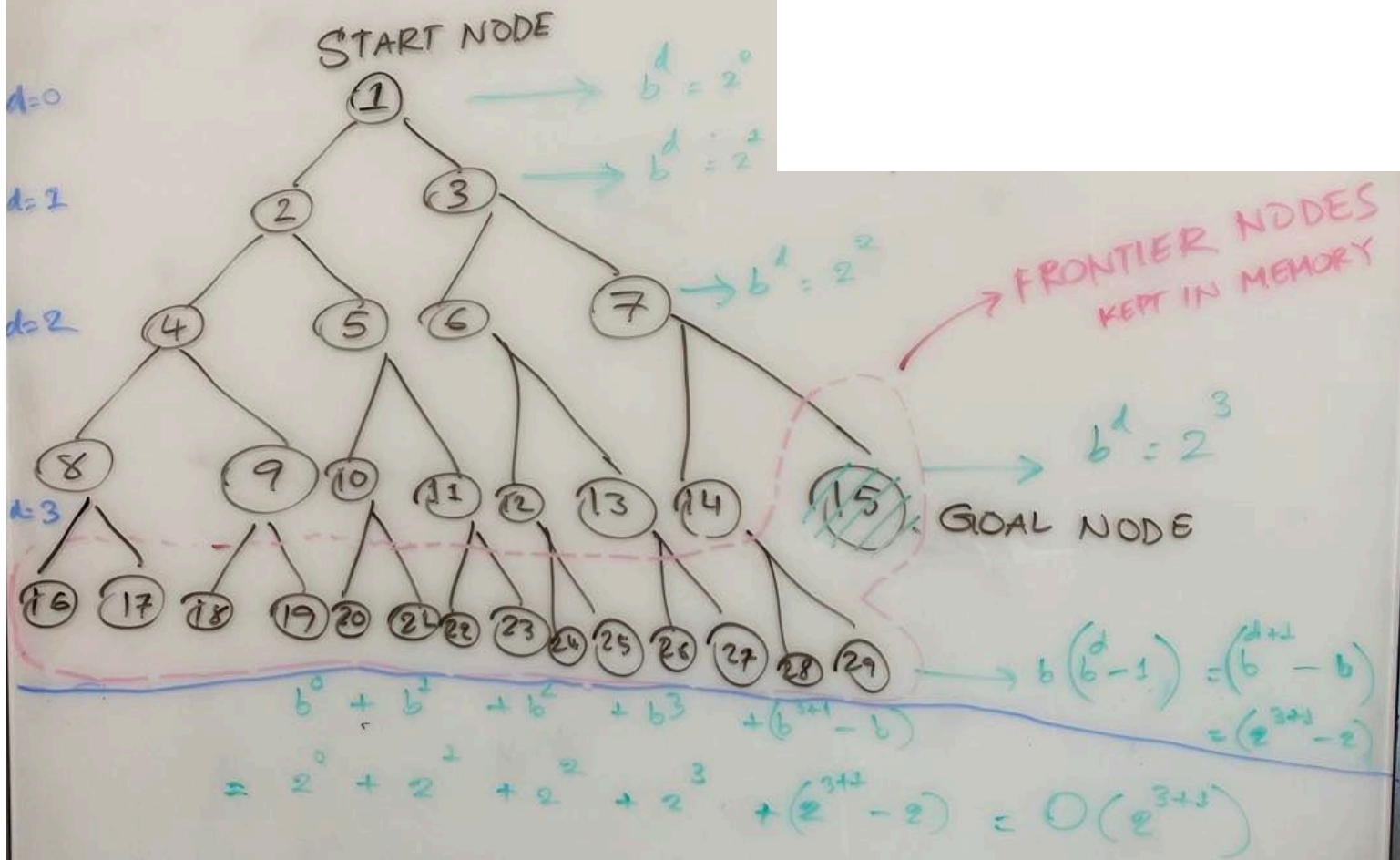
$$b \cdot (b^d - 1) = O(b^{d+1})$$

Time and space complexity is  $O(b^{d+1})$

- Memory is a big concern for BFS.
  - can easily generate nodes at 100MB/second so gigabytes of space required if BFS runs for hours

$b = 2$  : BRANCHING FACTOR  
 $d = 3$  : GOAL NODE AT THE SHALLOWEST DEPTH

TIME COMPLEXITY



# Properties of Breadth-First Search (BFS)

- Time is a big concern for BFS
  - If the problem has a solution at depth 16 with branching factor 10, it may take about 350 years for BFS to find it

<b>d</b>	<b># Nodes</b>	<b>Time</b>	<b>Memory</b>
<b>2</b>	111	0.11 msec	11 Kbytes
<b>4</b>	11,111	11.11 msec	1 Mbyte
<b>6</b>	$\sim 10^6$	1 sec	100 Mb
<b>8</b>	$\sim 10^8$	100 sec	10 Gbytes
<b>10</b>	$\sim 10^{10}$	2.8 hours	1 Tbyte
<b>12</b>	$\sim 10^{12}$	11.6 days	100 Tbytes
<b>14</b>	$\sim 10^{14}$	3.2 years	10,000 Tbytes

### Assumptions:

- $b = 10$
- A computer can process 1,000,000 nodes/sec

# New Content: Analysis of DFS

# Properties of Depth-First Search

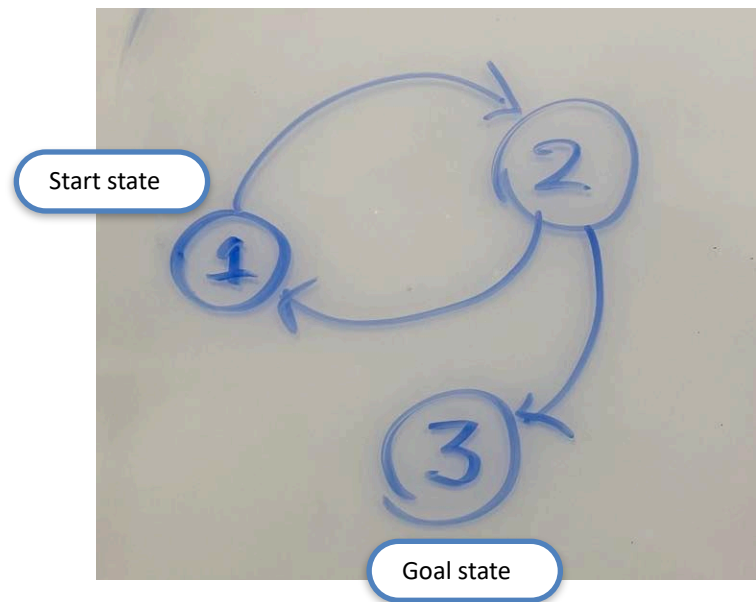
- **Completeness:**
  - a search algorithm is **complete** if it finds a solution whenever one exists
- **Optimality:**
  - a search algorithm is **optimal** if it returns a minimum-cost path if a solution exists
- **Time complexity:**
  - time required by the algorithm
- **Space complexity:**
  - memory required by the algorithm

# Properties of Depth-First Search

- Is DFS Complete?
- Is DFS Optimal?
- What is the running time complexity of DFS?
- What are the memory requirement of DFS?

# Properties of Depth-First Search

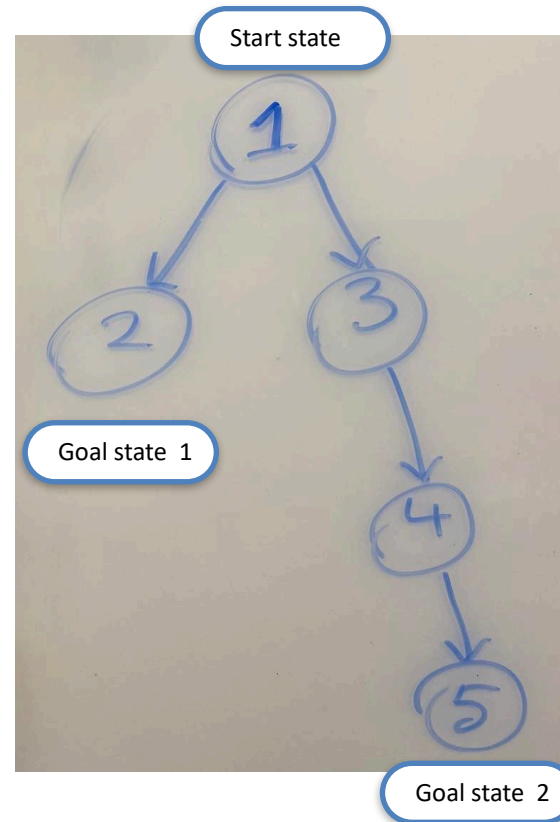
- Is DFS Complete?



- It is **not complete** even if the state graph is finite
  - The search tree could be infinite
  - Eg,  $[1] \rightarrow [2] \rightarrow [3, 1] \rightarrow [3, 2] \rightarrow [3, 3, 1] \rightarrow [3, 3, 2] \rightarrow \dots$

# Properties of Depth-First Search

- Is DFS Optimal?



- It is **not optimal**
  - The search tree could not reach goal state at a lower depth
  - Eg,  $[1] \rightarrow [2,3] \rightarrow [2,4] \rightarrow [2,5]$

# Properties of Depth-First Search

$b$ : branching factor

$l$ : depth of deepest leaf node

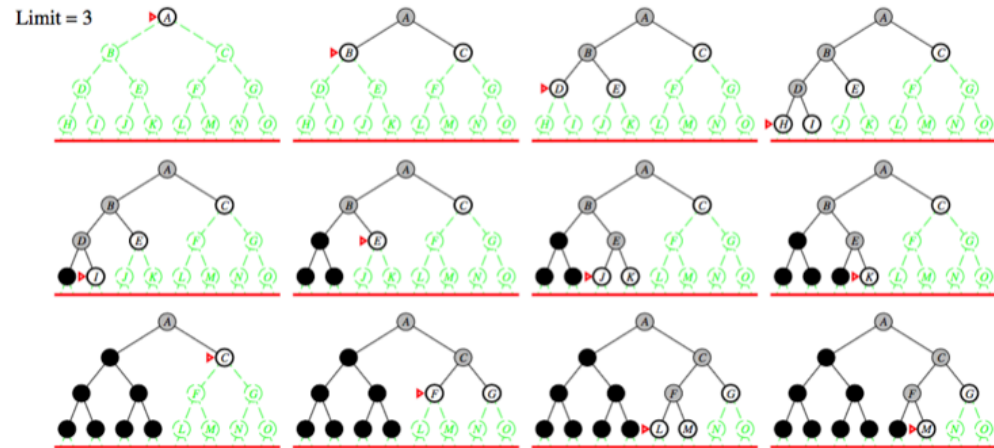
- Depth-first search
  - Not complete
  - Not optimal
- Number of nodes generated:  
 $1 + b + b^2 + \dots + b^l = O(b^l)$
- Time and space complexity is  $O(b \cdot l)$

# Analysis of IDS

# Properties of Iterative Deepening Search (IDS)

- Is IDS Complete?
- What is the running time complexity of IDS?
- What are the memory requirement of IDS?
- Is IDS Optimal?

# Properties of Iterative Deepening Search (IDS)



- IDS is complete
- IDS is optimal
- Number of nodes generated:

$$(d+1)b^0 + db^1 + (d-1)b^2 + \dots + b^d = O(b^d)$$

- Space complexity is  $O(b*d)$

# Number of Generated Nodes: BFS vs IDS

**b**: branching factor

**d**: depth of shallowest goal node

- **b** = 2 and **d** = 5

	BFS	IDS
Nodes at depth 1	1	$1 \times 6 = 6$
Nodes at depth 2	2	$2 \times 5 = 10$
Nodes at depth 3	4	$4 \times 4 = 16$
Nodes at depth 4	8	$8 \times 3 = 24$
Nodes at depth 5	16	$16 \times 2 = 32$
Nodes at depth 5	32	$32 \times 1 = 32$
	<b>63</b>	<b>120</b>

# Number of Generated Nodes: BFS vs IDS

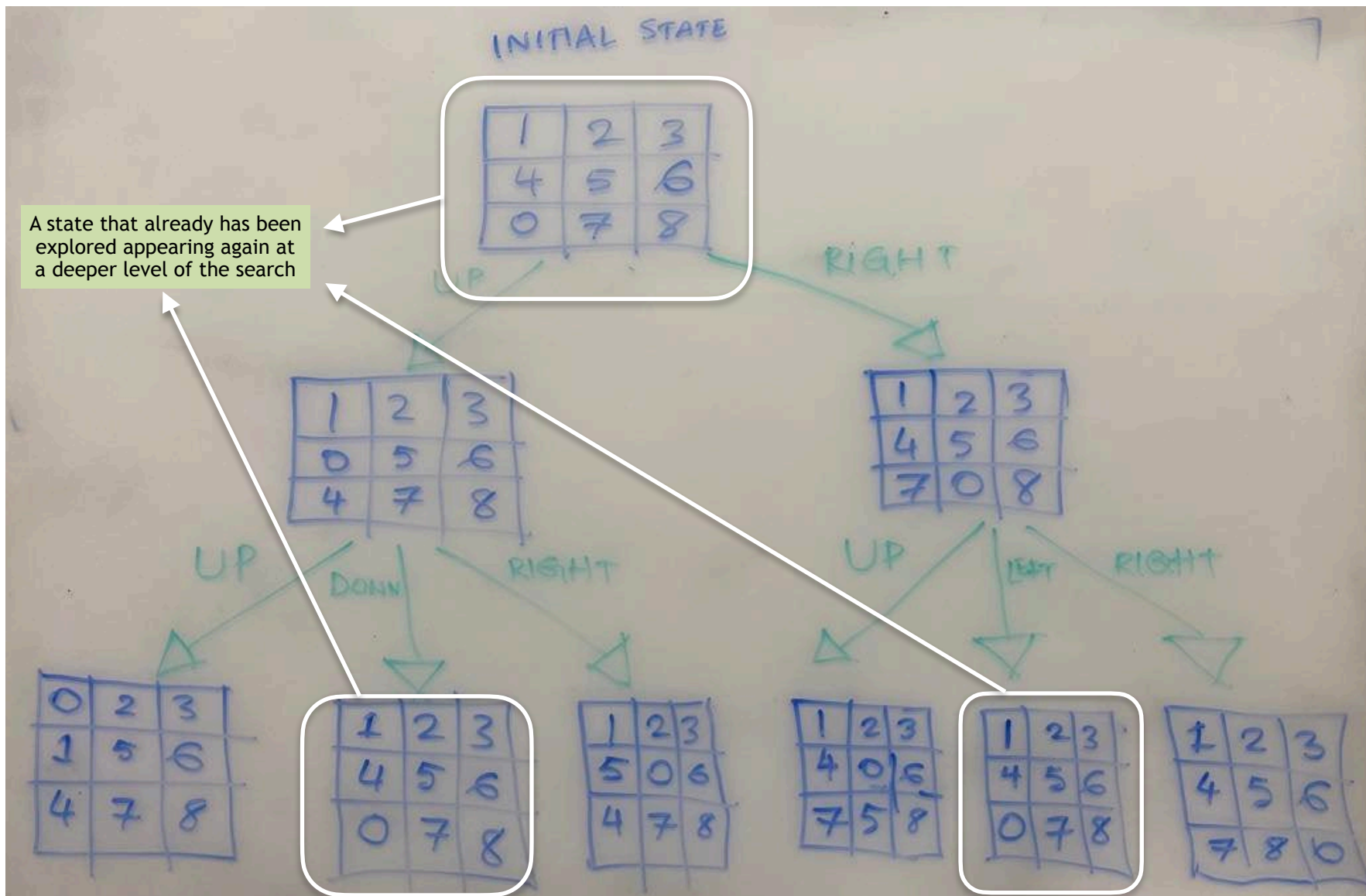
**b**: branching factor

**d**: depth of shallowest goal node

- **b** = 10 and **d** = 5

BFS	IDS
1	6
10	50
100	400
1,000	3,000
10,000	20,000
100,000	100,000
<b>111,111</b>	<b>123,456</b>

# Revisit 8-puzzle: Breadth-First Search



How can you stop revisiting states that have been explored?

# Avoiding Revisited States

- Requires comparing state descriptions
- Introduce a data-structure **EXPLORED**
  - Recall that we store unexpanded nodes in the **EXPLORED**
- For Breadth-First Search (BFS):
  - Store all states associated with generated nodes in **EXPLORED**
  - If a new node you found is in **EXPLORED**, then discard the node

# Python Set

- Python Set: “unordered collections of unique elements”
  - <https://docs.python.org/2/library/sets.html>
- Useful methods:
  - Given set `s` and an element `x`
    - Test for membership: `if x in s:`
    - Add an element: `s.add(x)`

# Python Set: example code

```
# ----- make a new set with three items -----  
print("Printing the current elements in the set: ", end="")  
my_set = set(['b', 'c', 'd'])           # make a new set with three items  
for elem in my_set:                     # iterate over the set elements  
    print(elem, end="")                 # note that a set doesn't maintain an order  
print("")
```

```
>>> %Run set_datastructure_example.py
```

```
Printing the current elements in the set: dbc
```

# Python Set: example code

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# ----- make a new set with three items -----
print("Printing the current elements in the set: ", end="")
my_set = set(['b', 'c', 'd'])           # make a new set with three items
for elem in my_set:                     # iterate over the set elements
    print(elem, end="")                 # note that a set doesn't maintain an order
print("")

# ----- add a new entry to the set -----
print("After adding 'e' and 'a' current elements in the set: ", end="")
my_set.add('e')                          # add a new entry
my_set.add('a')                          # add a new entry
print(my_set)                            # show the representation of the set
                                           # again note that a set doesn't maintain an
```

```
>>> %Run set_datastructure_example.py
```

```
Printing the current elements in the set: dbc
After adding 'e' and 'a' current elements in the set: {'b', 'a', 'd', 'c', 'e'}
```

# Python Set: example code

```
# ----- make a new set with three items -----
print("Printing the current elements in the set: ", end="")
my_set = set(['b', 'c', 'd'])          # make a new set with three items
for elem in my_set:                  # iterate over the set elements
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# ----- add a new entry to the set -----
print("After adding 'e' and 'a' current elements in the set: ", end="")
my_set.add('e')                       # add a new entry
my_set.add('a')                       # add a new entry
print(my_set)                         # show the representation of the set
# again note that a set doesn't maintain an

# ----- remove an entry from the set -----
elem = my_set.pop()                  # return and remove an arbitrary element
print("Popped element (randomly) from the set: ", elem)

print("The set that remains is: ", end="")
print(my_set)
```

```
>>> %Run set_datastructure_example.py
```

```
Printing the current elements in the set: dbc
After adding 'e' and 'a' current elements in the set: {'b', 'a', 'd', 'c', 'e'}
Popped element (randomly) from the set: b
The set that remains is: {'a', 'd', 'c', 'e'}
```

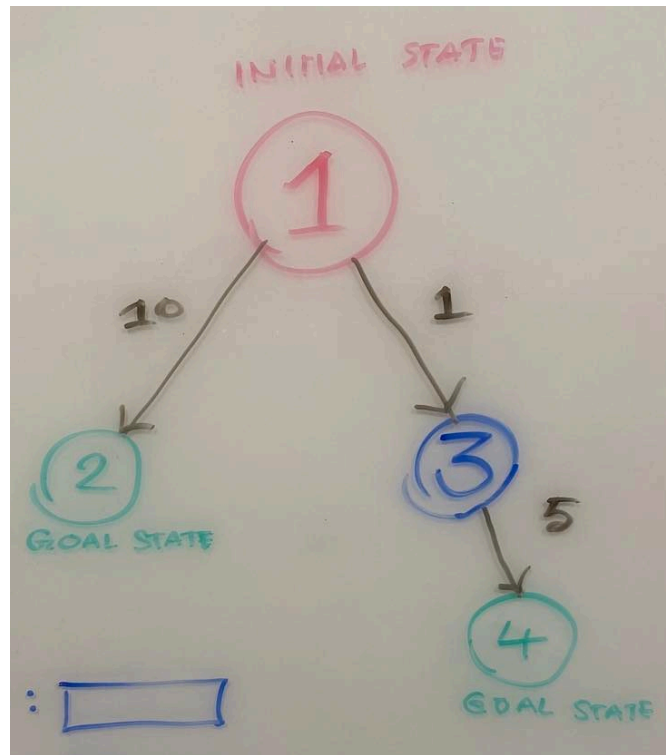
# Python Set: example code

```
# ----- test membership in the set -----  
print("Testing membership in the set: ", end="")  
elem = 'c'  
if elem in my_set:  
    print (elem, " is in the set")  
else:  
    print(elem, " is not in the set")
```

```
Testing membership in the set: c is in the set
```

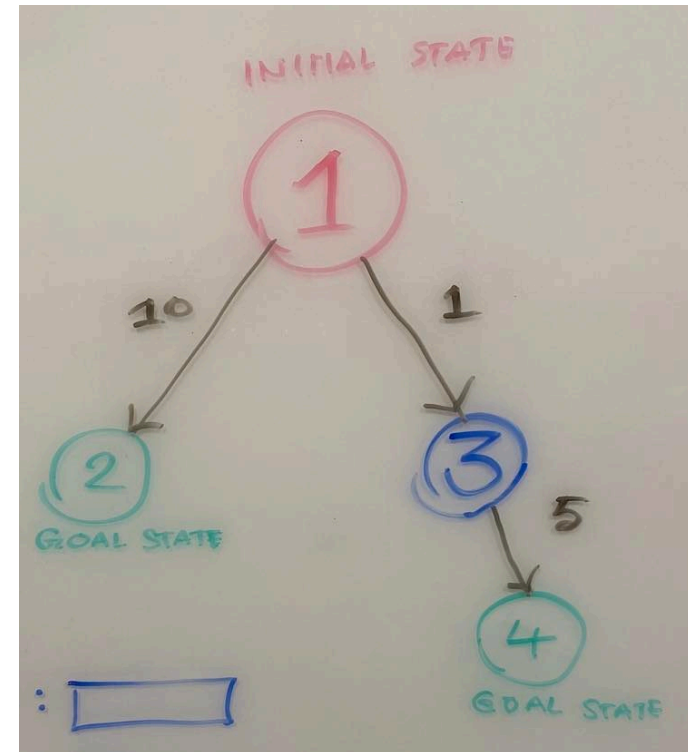
# Non-uniform cost function

- Lets add a cost to each edge in the state graph

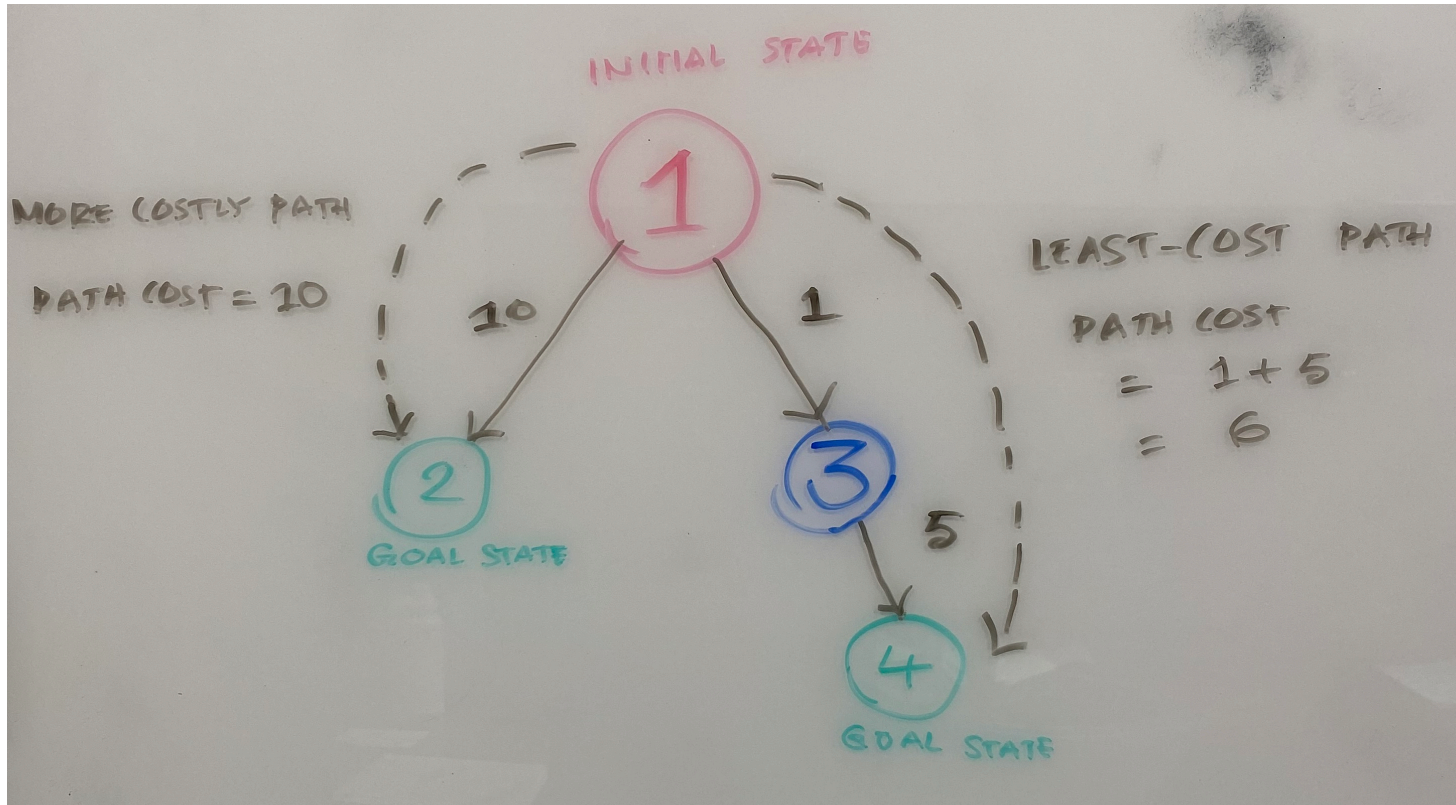


# Non-uniform cost function

- What is the optimal cost path in this state graph?
- What will BFS do in this state graph?
- Can BFS reach a goal node with an optimal cost?

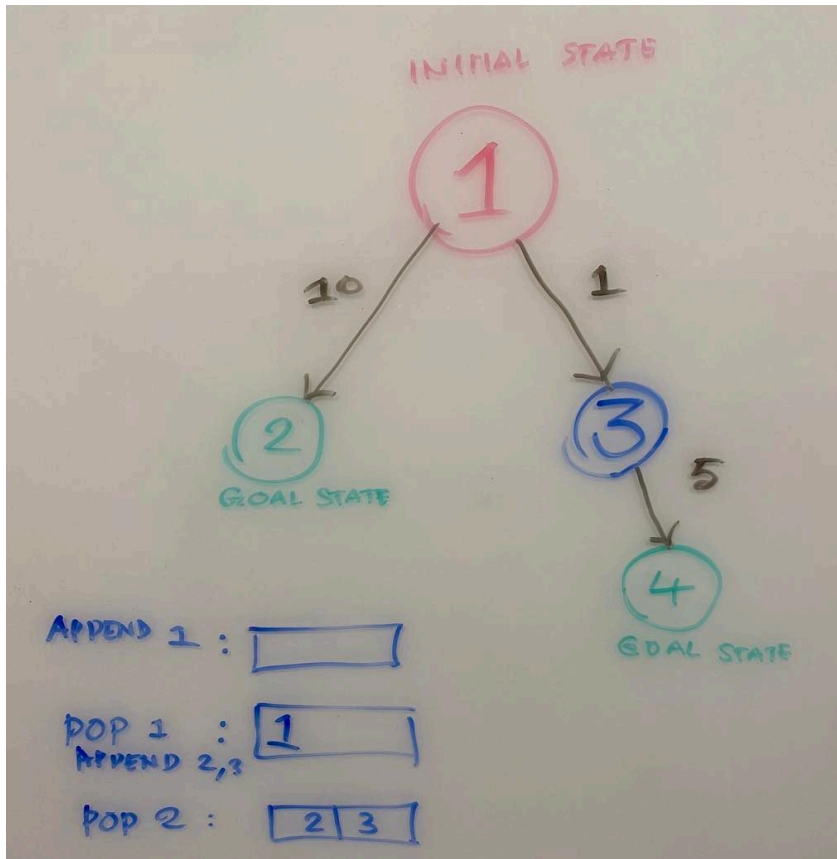


# Non-uniform cost function

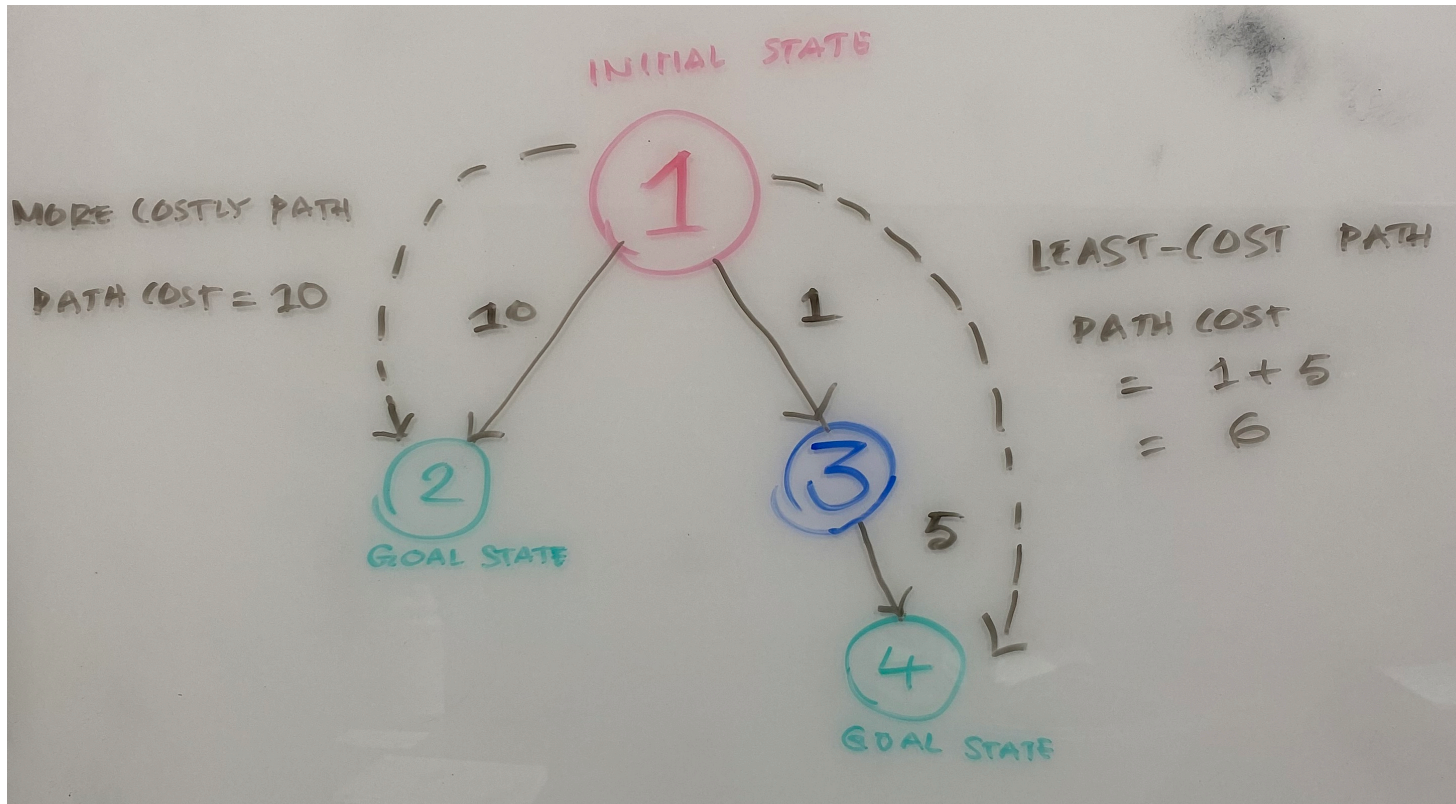


# Non-uniform cost function

- BFS is no longer optimal
  - when BFS reaches node 2; stops the search



# Non-uniform cost function



- BFS will find the more costly path hence **not optimal**

# Non-uniform cost function

- Since BFS is no longer optimal, so what can we do:
  - instead of expanding the node in FRINGE with the shortest path from the initial state, we want to the one with the **least costly** path

# Uniform Cost Search

- If we keep track of a “cost” for expanding a node, then we desire an optimal solution with a minimal total path cost
- The node from the frontier that is selected to be expanded is the node that has the minimum total path cost
  - thus the frontier is implemented as a *priority queue*

# Uniform Cost Search

- pseudo-code from Russell & Norvig textbook

```
function UNIFORM-COST-SEARCH(problem) returns a solution, or failure
  node ← a node with STATE = problem.INITIAL-STATE, PATH-COST = 0
  frontier ← a priority queue ordered by PATH-COST, with node as the only element
  explored ← an empty set
  loop do
    if EMPTY?(frontier) then return failure
    node ← POP(frontier) /* chooses the lowest-cost node in frontier */
    if problem.GOAL-TEST(node.STATE) then return SOLUTION(node)
    add node.STATE to explored
    for each action in problem.ACTIONS(node.STATE) do
      child ← CHILD-NODE(problem, node, action)
      if child.STATE is not in explored or frontier then
        frontier ← INSERT(child, frontier)
      else if child.STATE is in frontier with higher PATH-COST then
        replace that frontier node with child
```

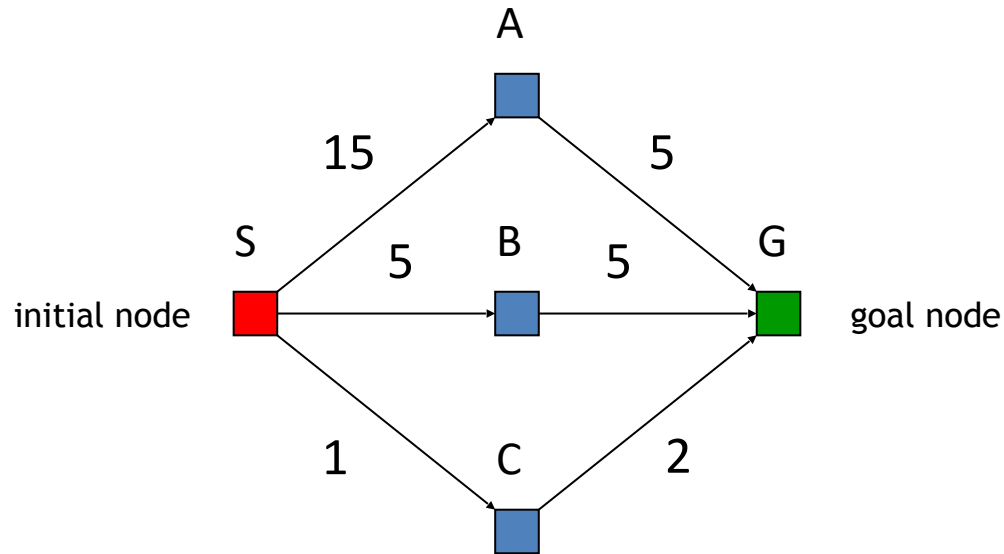
use Priority Queue  
instead of deque

# Uniform Cost Search

- Uniform Cost Search is essentially graph search algorithm
  - It terminates if goal state is *removed* from frontier (priority queue)
  - Checks if a shorter path to a current node on the frontier is possible

# Priority Queue to the rescue

- Can “Uniform Cost Search” (with **FRONTIER** implemented using Priority Queue) reach a goal node with an optimal cost?



# Priority Queue to the rescue

- Search tree with node expansions are as follows:
  - “Uniform Cost Search” (with **FRONTIER** implemented using Priority Queue) reaches a goal node with an optimal cost

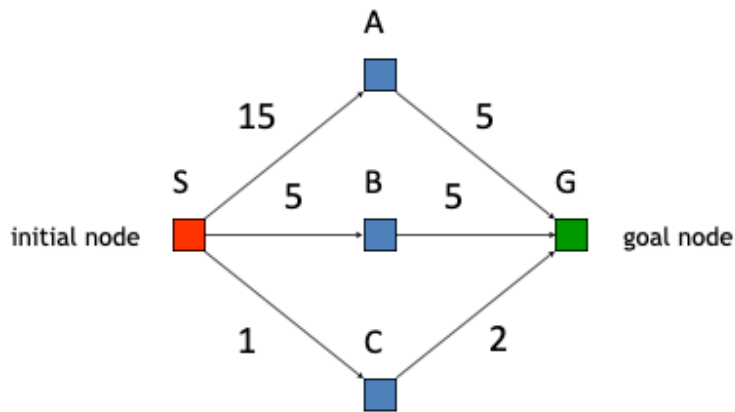


Figure: State Graph

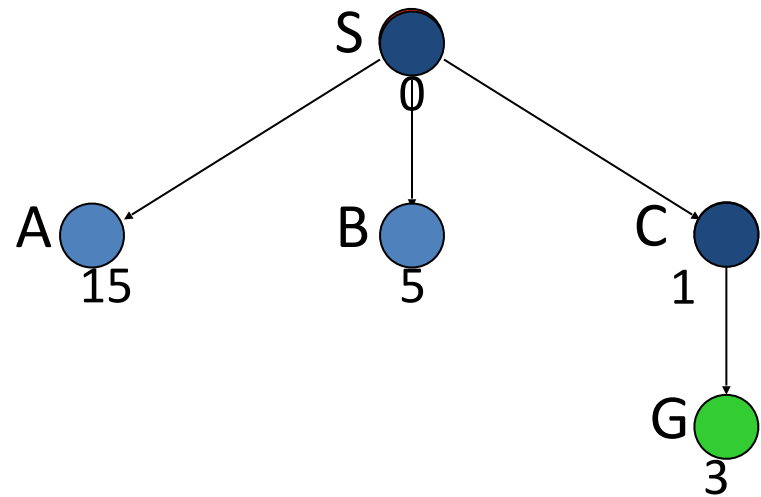


Figure: Order of node expansions

# Categorizing Search Strategies

- Search algorithms we have seen so far
  - Breadth-First Search (BFS)
  - Depth-First Search (DFS)
  - Depth-Limited Search (DLS)
  - Iterative Deepening Search (IDS)
  - Uniform Cost Search (UCS)