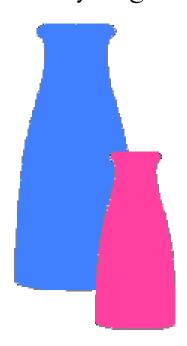
State-Space Searches

State spaces

- A state space consists of
 - A (possibly infinite) set of states
 - The start state represents the initial problem
 - Each state represents some configuration reachable from the start state
 - Some states may be goal states (solutions)
 - A set of operators
 - Applying an operator to a state transforms it to another state in the state space
 - Not all operators are applicable to all states
- State spaces are used extensively in Artificial Intelligence (AI)

"You are given two jugs, a 4-litre one and a 3-litre one. Neither has any measuring markers on it. There is a pump that can be used to fill the jugs with water. How can you get exactly 2 litres of water into 4-litre jug."



• State: (x, y)

$$x = 0, 1, 2, 3, \text{ or } 4$$
 $y = 0, 1, 2, 3$

- Start state: (0, 0).
- Goal state: (2, n) for any n.
- Attempting to end up in a goal state.

- 1. (x, y) $\rightarrow (4, y)$ if x < 4
- 2. $(x, y) \rightarrow (x, 3)$ if y < 3
- 3. (x, y) $\rightarrow (x d, y)$ if x > 0
- 4. (x, y) $\rightarrow (x, y d)$ if y > 0

```
5. \quad (\mathbf{x}, \mathbf{y})
                                    \rightarrow (0, y)
       if x > 0
6. (x, y)
                                    \rightarrow (x, 0)
      if y > 0
7. (x, y)
                                    \rightarrow (4, y - (4 - x))
       if x + y \ge 4, y > 0
8. \quad (\mathbf{x}, \mathbf{y})
                             \rightarrow (x - (3 - y), 3)
      if x + y \ge 3, x > 0
```

9.
$$(x, y) \rightarrow (x + y, 0)$$

if $x + y \le 4$, $y > 0$
10. $(x, y) \rightarrow (0, x + y)$
if $x + y \le 3$, $x > 0$
11. $(0, 2) \rightarrow (2, 0)$
12. $(2, y) \rightarrow (0, y)$

State Space Search: Water Jug Problem

- 1. current state = (0, 0)
- 2. Loop until reaching the goal state (2, 0)
 - Apply a rule whose left side matches the current state
 - Set the new current state to be the resulting state
 - (0,0)
 - (0, 3)
 - (3, 0)
 - (3, 3)
 - (4, 2)
 - (0, 2)
 - (2, 0)

Example 2: The 15-puzzle

Start state:

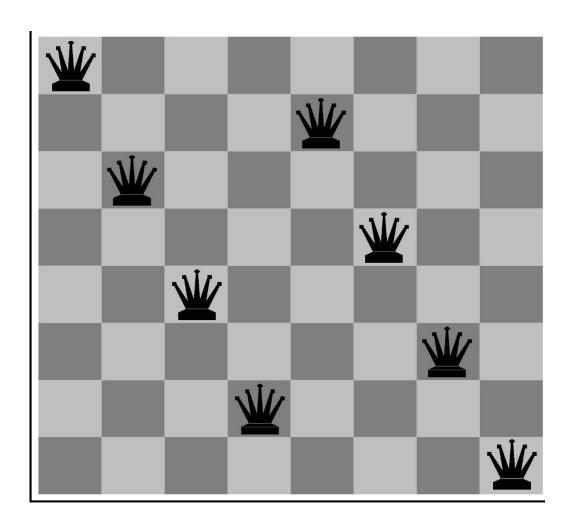
3	10	13	7
9	14	6	1
4		15	2
11	8	5	12

Goal state:

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

- The start state is some (almost) random configuration of the tiles
- The goal state is as shown
- Operators are
 - Move empty space up
 - Move empty space down
 - Move empty space right
 - Move empty space left
- Operators apply if not against edge

8 Queen Problem



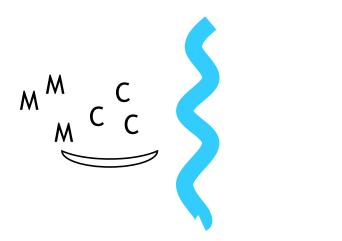
State-Space problem formulation

- states? -any arrangement of n<=8 queens
 -or arrangements of n<=8 queens in leftmost n
 columns, 1 per column, such that no queen
 attacks any other.
- <u>initial state?</u> no queens on the board
- actions? -add queen to any empty square
 -or add queen to leftmost empty square such that it is not attacked by other queens.
- goal test? 8 queens on the board, none attacked.
- path cost? 1 per move

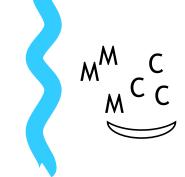
Example 3: Missionaries and cannibals

• An old puzzle is the "Missionaries and cannibals" problem (in various guises)

- The missionaries and cannibals wish to cross a river
- They have a canoe that can hold two people
- It is unsafe to have cannibals outnumber missionaries







Goal state



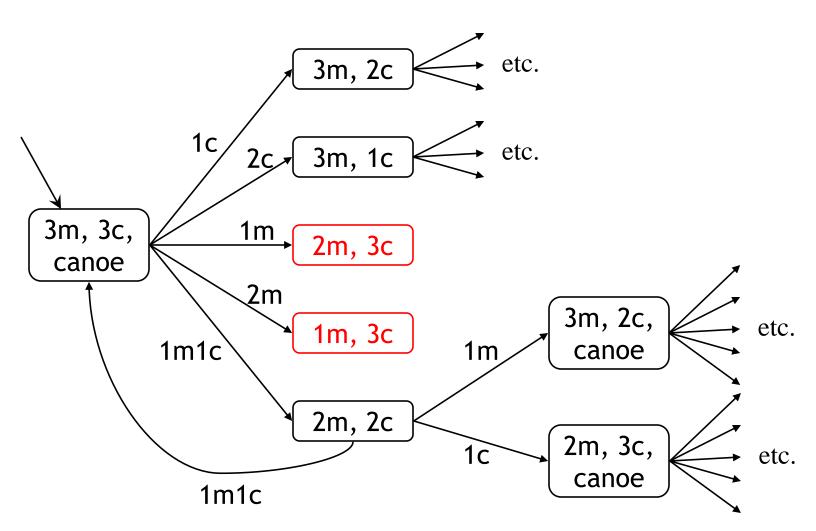
States

- A *state* can be represented by the number of missionaries and cannibals on each side of the river
 - Initial state: 3m,3c,canoe / 0m,0c
 - Goal state: 0m,0c / 3m,3c,canoe
 - We assume that crossing the river is a simple procedure that always works (so we don't have to represent the canoe being in the middle of the river)
- However, this is redundant; we only need to represent how many missionaries/cannibals are on *one* side of the river
 - Initial state: 3m,3c,canoe
 - Goal state: 0m,0c

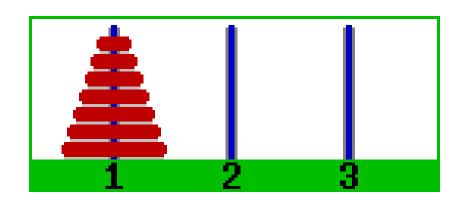
Operations

- An *operation* takes us from one state to another
- Here are five possible operations:
 - Canoe takes 1 missionary across river (1m)
 - Canoe takes 1 cannibal across river (1C)
 - Canoe takes 2 missionaries across river (2m)
 - Canoe takes 2 cannibals across river (2c)
 - Canoe takes 1 missionary and 1 cannibal across river (1m1c)
- We don't have to specify "west to east" or "east to west" because only one of these will be possible at any given time

The state space



Example Problems - Towers of Hanoi



States: combinations of poles and disks

Operators: move disk x from pole y to pole z subject to constraints

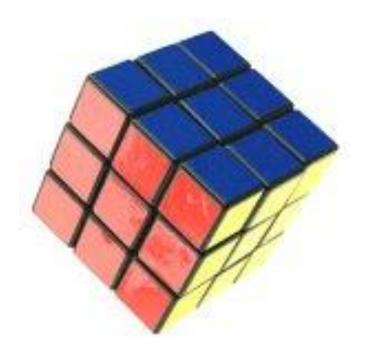
- cannot move disk on top of smaller disk
- cannot move disk if other disks on top

Goal test: disks from largest (at bottom) to smallest on goal pole

Path cost: 1 per move

Towers of Hanoi applet

Example Problems - Rubik's Cube



States: list of colors for each cell on each face

Initial state: one specific cube configuration

Operators: rotate row x or column y on face z direction a

Goal: configuration has only one color on each face

Path cost: 1 per move

Rubik's cube applet

State-space searching

- Most problems in AI can be cast as searches on a state space
- The space can be tree-shaped or graph-shaped
 - If a graph, need some way to keep track of where you have been, so as to avoid loops
- The state space is often very, very large
- We can minimize the size of the search space by careful choice of operators
- Exhaustive searches don't work—we need *heuristics*

Sample Search Problems

- Graph coloring
- Protein folding
- Game playing
- Airline travel
- Proving algebraic equalities
- Robot motion planning

The basic search algorithm

Initialize: put the start node into OPEN while OPEN is not empty take a node N from OPEN if N is a goal node, report success put the children of N onto OPEN Report failure

- If OPEN is a stack, this is a depth-first search
- If OPEN is a queue, this is a breadth-first search
- If OPEN is a priority queue, sorted according to most promising first, we have a best-first search

Uninformed Search

A General State-Space Search Algorithm

- Node n
 - state description
 - parent (may use a backpointer) (if needed)
 - Operator used to generate n (optional)
 - Depth of n (optional)
 - Path cost from S to n (if available)
- OPEN list
 - initialization: {S}
 - node insertion/removal depends on specific search strategy
- CLOSED list
 - initialization: {}
 - organized by backpointers

A General State-Space Search Algorithm

```
open := \{S\}; closed :=\{\};
repeat
  n := select(open); /* select one node from open for expansion */
   if n is a goal
      then exit with success; /* delayed goal testing */
   expand(n)
          /* generate all children of n
            put these newly generated nodes in open (check duplicates)
            put n in closed (check duplicates) */
until open = \{\};
exit with failure
```

Some Issues

- Search process constructs a search tree, where
 - **root** is the initial state S, and
 - leaf nodes are nodes
 - not yet been expanded (i.e., they are in OPEN list) or
 - having no successors (i.e., they're "deadends")
- Search tree may be infinite because of loops even if state space is small
- Search strategies mainly differ on select (open)
- Each node represents a partial solution path (and cost of the partial solution path) from the start node to the given node.
 - in general, from this node there are many possible paths (and therefore solutions) that have this partial path as a prefix.

Evaluating Search Strategies

Completeness

• Guarantees finding a solution whenever one exists

Time Complexity

• How long (worst or average case) does it take to find a solution? Usually measured in terms of the **number of nodes expanded**

Space Complexity

• How much space is used by the algorithm? Usually measured in terms of the maximum size that the "OPEN" list becomes during the search

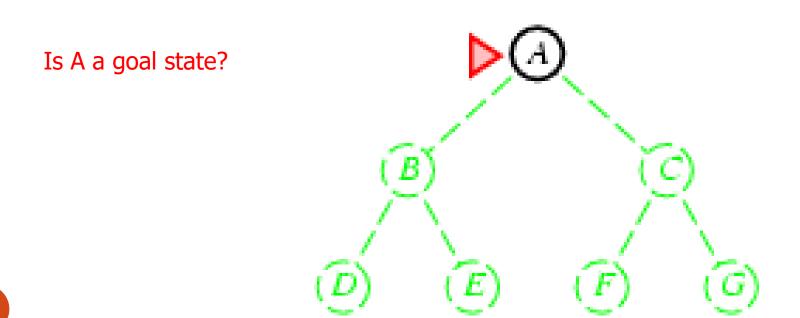
Optimality/Admissibility

• If a solution is found, is it guaranteed to be an optimal one? For example, is it the one with minimum cost?

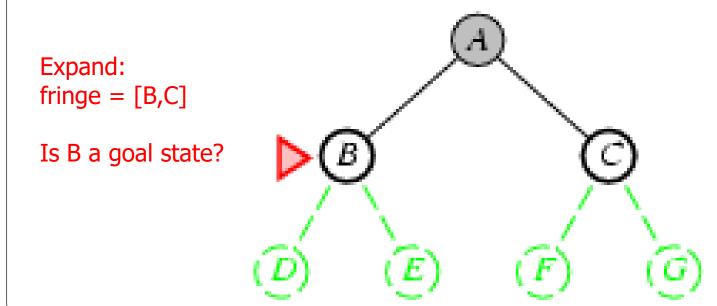
Uninformed search strategies

- Uninformed: While searching you have no clue whether one non-goal state is better than any other. Your search is blind.
- Various blind strategies:
- Breadth-first search
- Uniform-cost search
- Depth-first search
- Iterative deepening search

- Expand shallowest unexpanded node
- Implementation:
 - *fringe* is a first-in-first-out (FIFO) queue, i.e., new successors go at end of the queue.



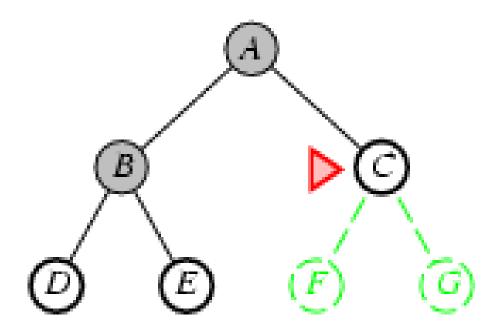
- Expand shallowest unexpanded node
- Implementation:
 - fringe is a FIFO queue, i.e., new successors go at end



- Expand shallowest unexpanded node
- Implementation:
 - fringe is a FIFO queue, i.e., new successors go at end

Expand: fringe=[C,D,E]

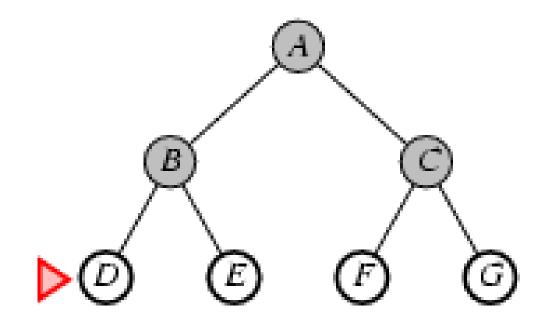
Is C a goal state?



- Expand shallowest unexpanded node
- Implementation:
 - fringe is a FIFO queue, i.e., new successors go at end

Expand: fringe=[D,E,F,G]

Is D a goal state?

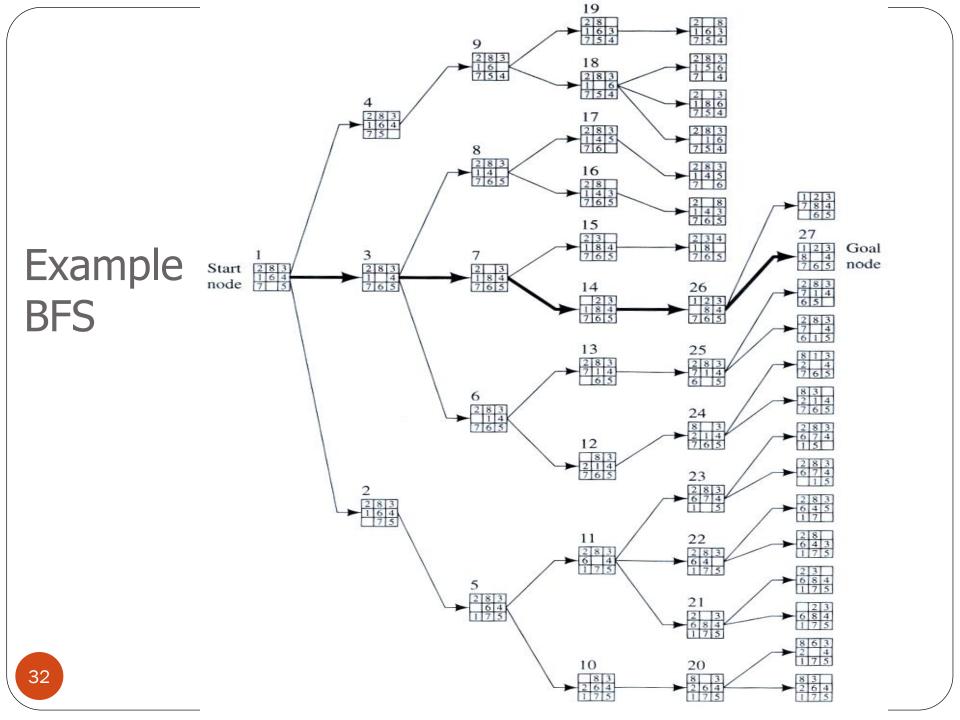


8 Puzzel Game

2	8	3
1	6	4
7		5

2		8
1	6	3
7	5	4

Initial State Goal State



Analysis

- See what happens with b=10
 - expand 10,000 nodes/second
 - 1,000 bytes/node

Depth	Nodes	Time	Memory
2	1110	.11 seconds	1 megabyte
4	111,100	11 seconds	106 megabytes
6	10^{7}	19 minutes	10 gigabytes
8	10^{9}	31 hours	1 terabyte
10	10 ¹¹	129 days	101 terabytes
12	10^{13}	35 years	10 petabytes
15	10^{15}	3,523 years	1 exabyte

Properties of breadth-first search

- <u>Complete?</u> Yes it always reaches goal (if *b* is finite)
- <u>Time?</u> $1+b+b^2+b^3+...+b^d+(b^{d+1}-b)) = O(b^{d+1})$ (this is the number of nodes we generate)
- Space? $O(b^{d+1})$ (keeps every node in memory, either in fringe or on a path to fringe).
- Optimal? Yes (if we guarantee that deeper solutions are less optimal, e.g. step-cost=1).
- Space is the bigger problem (more than time)

Uniform-cost search

Breadth-first is only optimal if step costs is increasing with depth (e.g. constant). Can we guarantee optimality for any step cost?

Uniform-cost Search: Expand node with smallest path cost g(n).

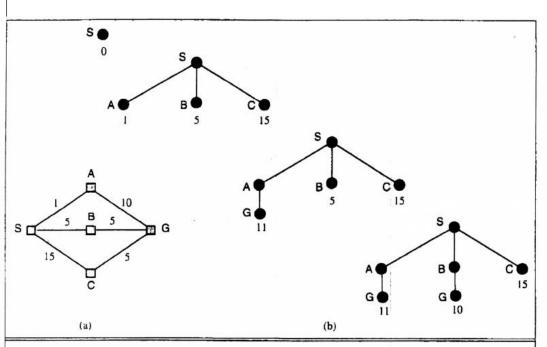
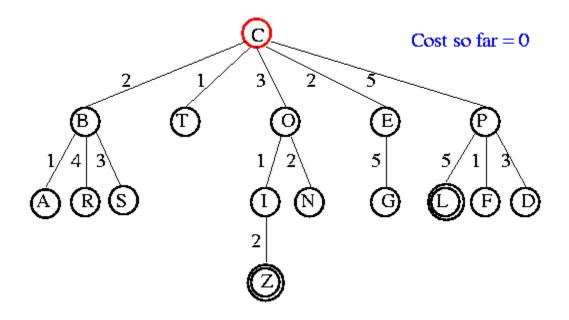
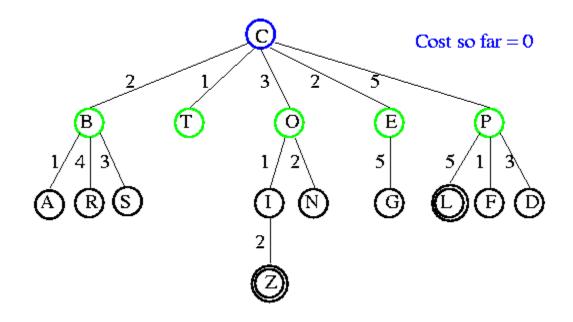


Figure 3.13 A route-finding problem. (a) The state space, showing the cost for each operator. (b) Progression of the search. Each node is labelled with g(n). At the next step, the goal node with g = 10 will be selected.

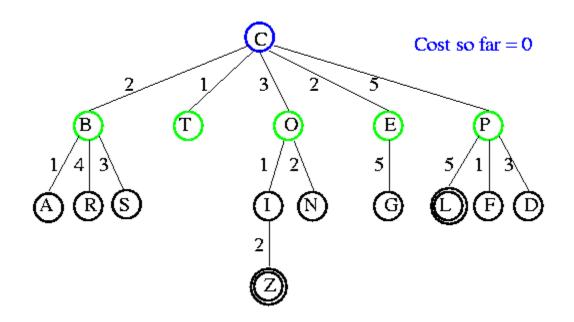
UCS Example



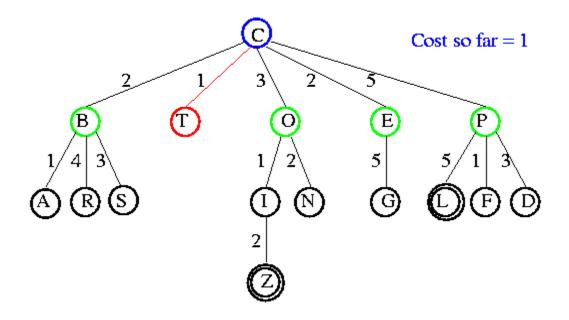
Open list: C



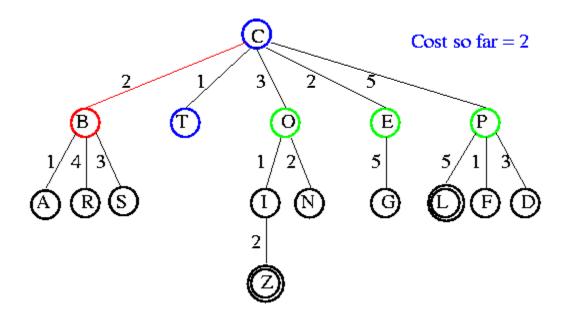
Open list: B(2) T(1) O(3) E(2) P(5)



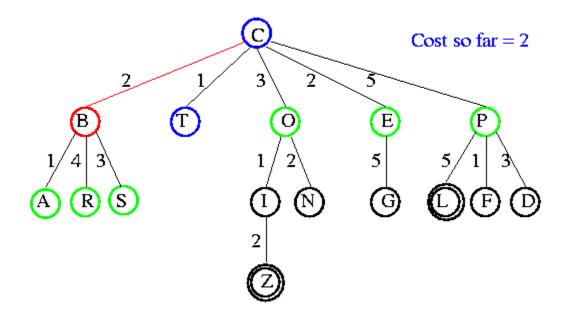
Open list: T(1) B(2) E(2) O(3) P(5)



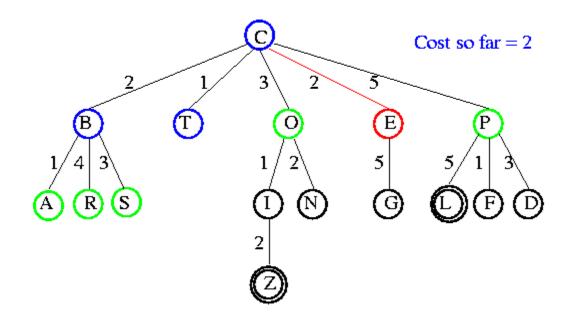
Open list: B(2) E(2) O(3) P(5)



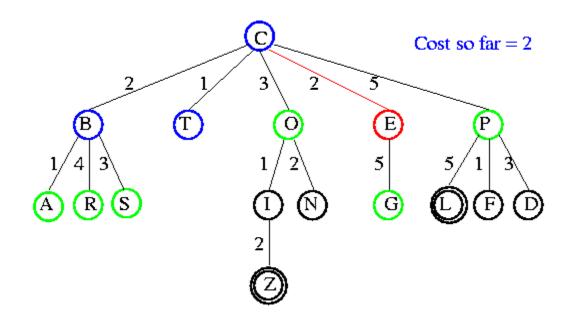
Open list: E(2) O(3) P(5)



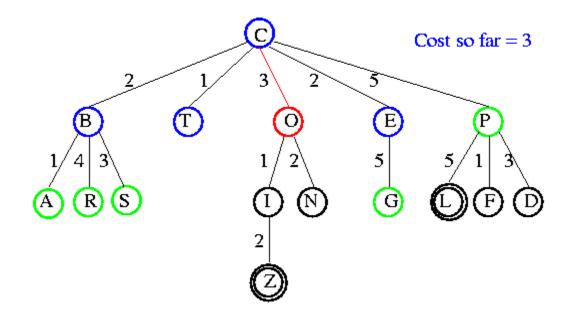
Open list: E(2) O(3) A(3) S(5) P(5) R(6)



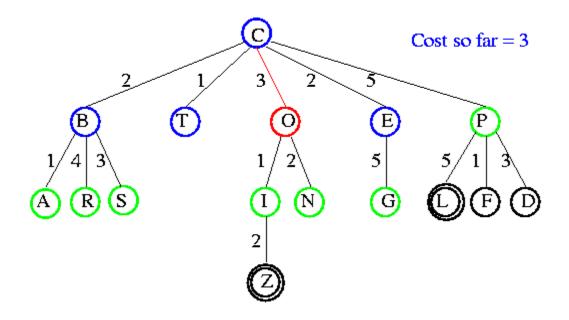
Open list: O(3) A(3) S(5) P(5) R(6)



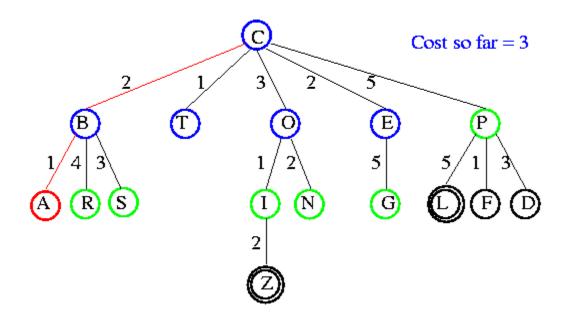
Open list: O(3) A(3) S(5) P(5) R(6) G(10)



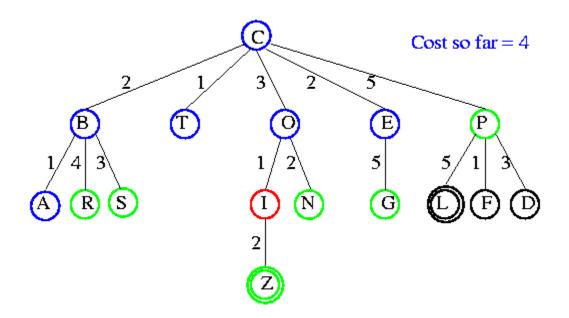
Open list: A(3) S(5) P(5) R(6) G(10)



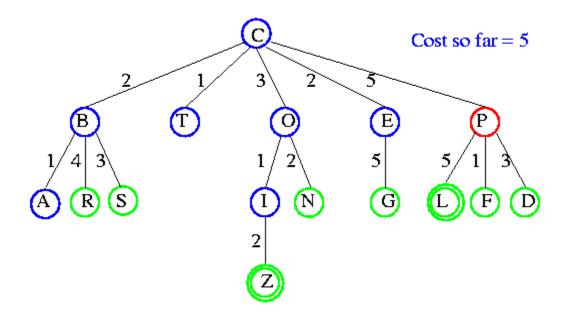
Open list: A(3) I(4) S(5) N(5) P(5) R(6) G(10)



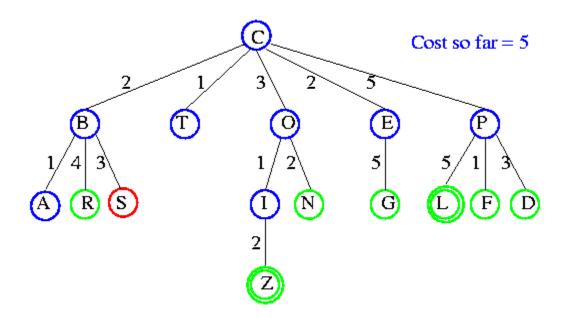
Open list: I(4) P(5) S(5) N(5) R(6) G(10)



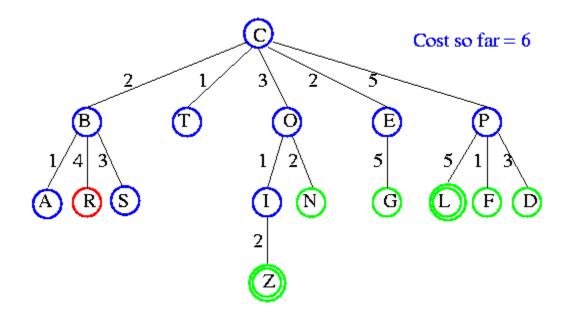
Open list: P(5) S(5) N(5) R(6) Z(6) G(10)



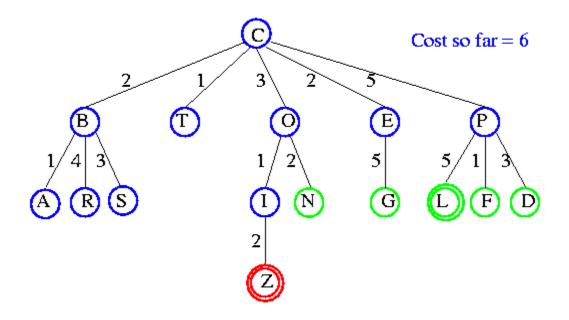
Open list: S(5) N(5) R(6) Z(6) F(6) D(8) G(10) L(10)



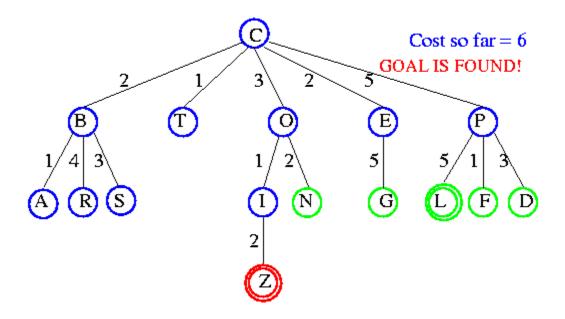
Open list: N(5) R(6) Z(6) F(6) D(8) G(10) L(10)



Open list: Z(6) F(6) D(8) G(10) L(10)



Open list: F(6) D(8) G(10) L(10)



Uniform-cost search

Implementation: fringe = queue ordered by path cost Equivalent to breadth-first if all step costs all equal.

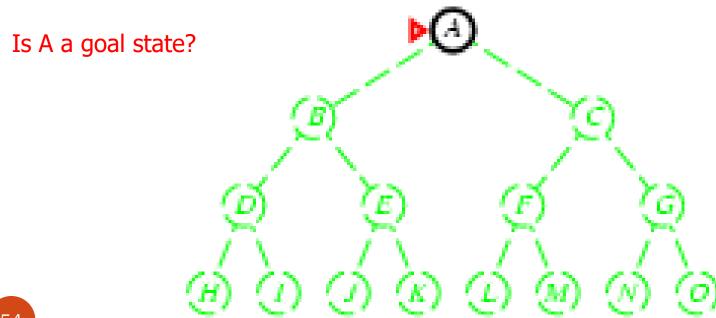
Complete? Yes, if step cost ≥ ε (otherwise it can get stuck in infinite loops)

<u>Time?</u> # of nodes with $path cost \leq cost of optimal solution.$

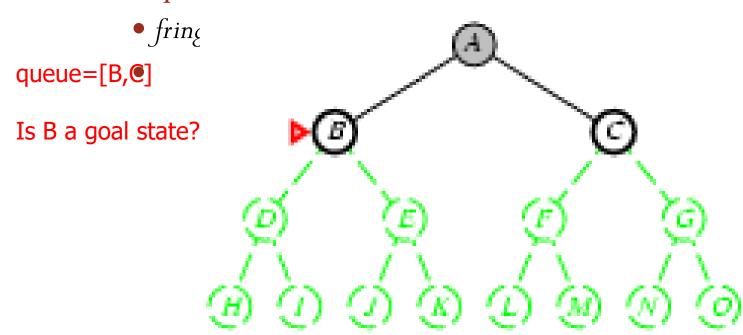
<u>Space?</u> # of nodes on paths with path cost ≤ cost of optimal solution.

Optimal? Yes, for any step cost.

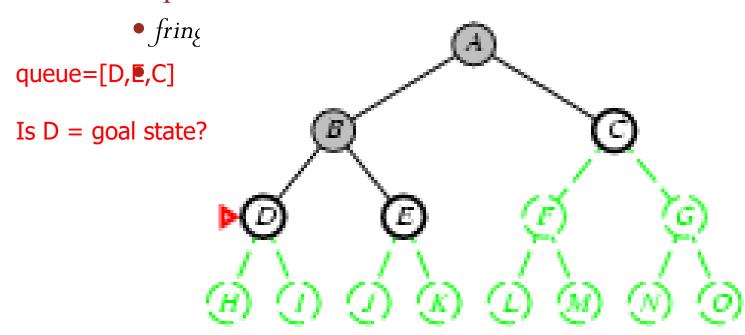
- Expand deepest unexpanded node
- Implementation:
 - fringe = Last In First Out (LIPO) queue, i.e., put successors at front



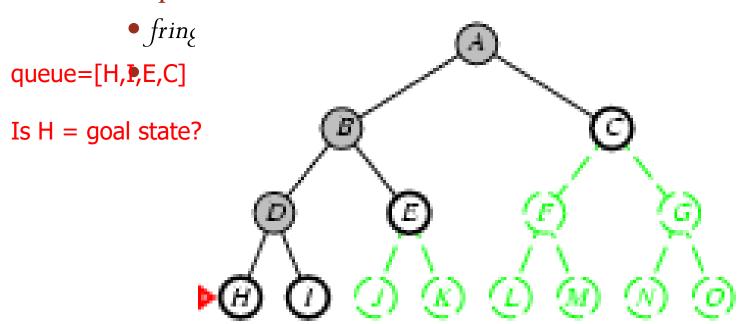
- Expand deepest unexpanded node
- Implementation:



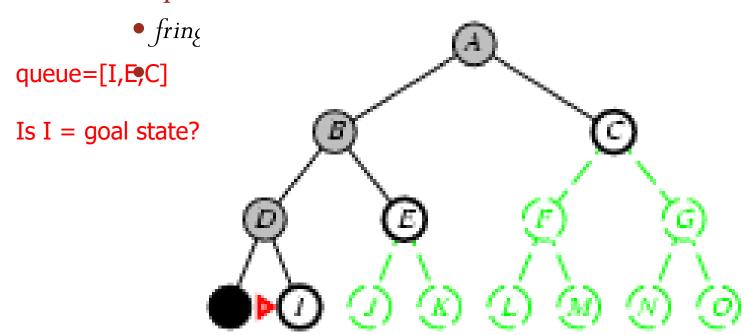
- Expand deepest unexpanded node
- Implementation:



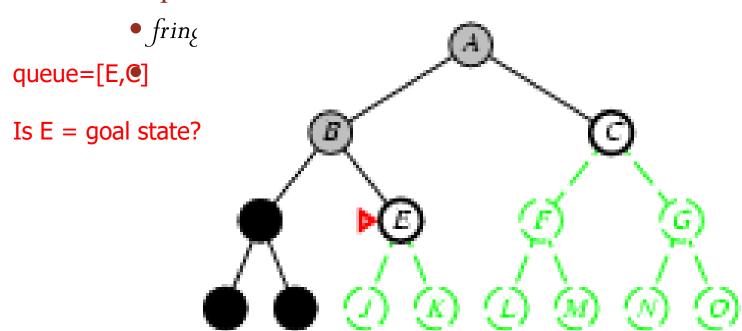
- Expand deepest unexpanded node
- Implementation:



- Expand deepest unexpanded node
- Implementation:



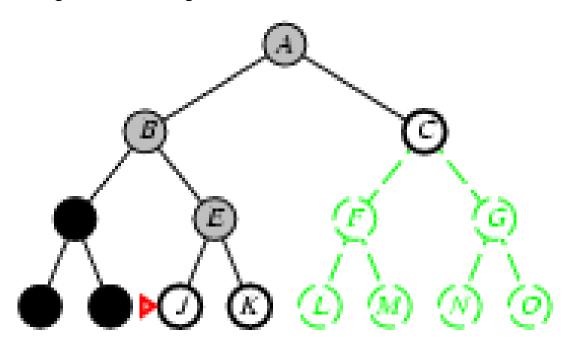
- Expand deepest unexpanded node
- Implementation:



- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front

queue=[J,K,C]

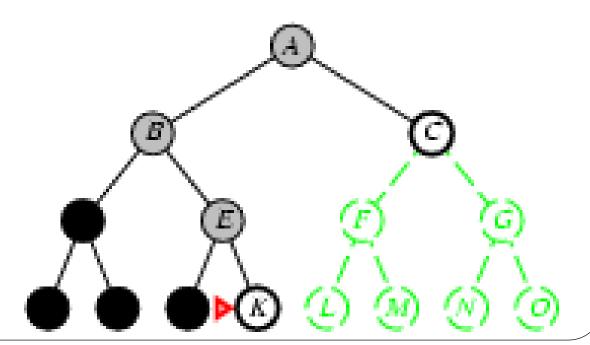
Is J = goal state?



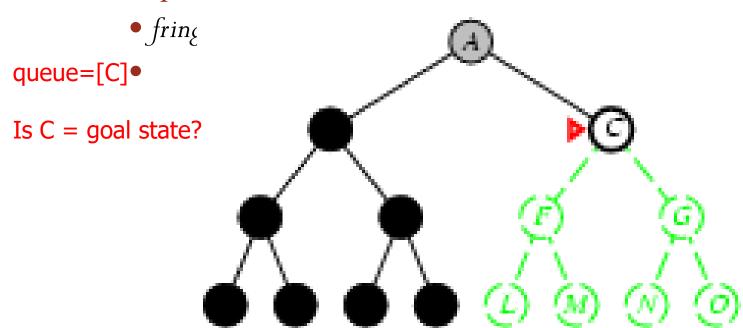
- Expand deepest unexpanded node
- Implementation:
 - fringe = LIFO queue, i.e., put successors at front

queue=[K,€]

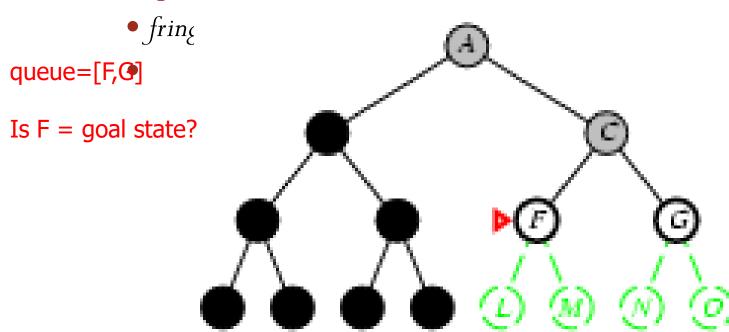
Is K = goal state?



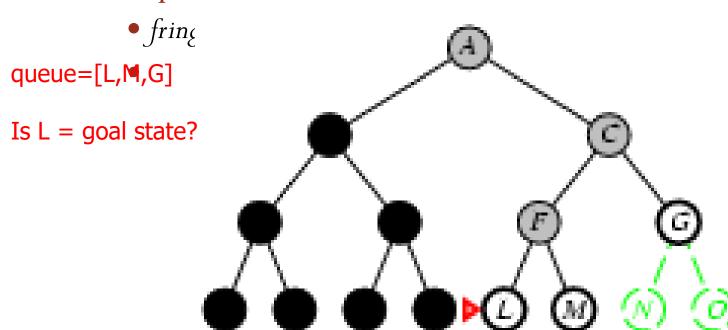
- Expand deepest unexpanded node
- Implementation:



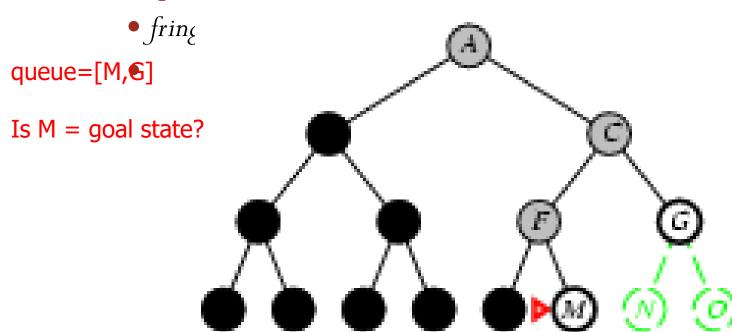
- Expand deepest unexpanded node
- Implementation:



- Expand deepest unexpanded node
- Implementation:



- Expand deepest unexpanded node
- Implementation:



Properties of depth-first search

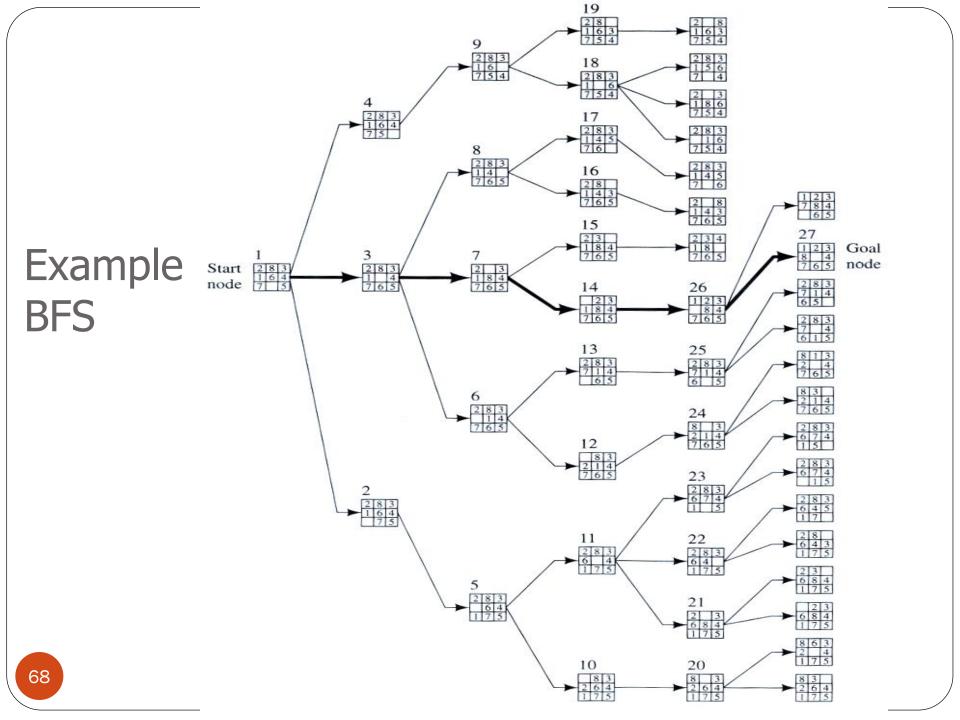
- <u>Complete?</u> No: fails in infinite-depth spaces
 Can modify to avoid repeated states along path
- Time? $O(b^m)$ with m=maximum depth
- terrible if *m* is much larger than *d*
 - but if solutions are dense, may be much faster than breadth-first
- <u>Space?</u> O(bm), i.e., linear space! (we only need to remember a single path + expanded unexplored nodes)
- Optimal? No (It may find a non-optimal goal first)

8 Puzzel Game

2	8	3
1	6	4
7		5

8		3
2	6	4
1	7	5

Initial State Goal State



Iterative deepening search

- To avoid the infinite depth problem of DFS, we can decide to only search until depth L, i.e. we don't expand beyond depth L.
 - → Depth-Limited Search
- What of solution is deeper than L? → Increase L iteratively.
 - → Iterative Deepening Search
- As we shall see: this inherits the memory advantage of Depth-First search.

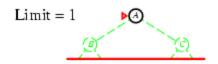
Iterative deepening search *L*=0

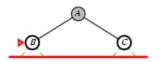
Limit = 0

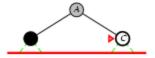


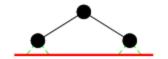


Iterative deepening search L=1

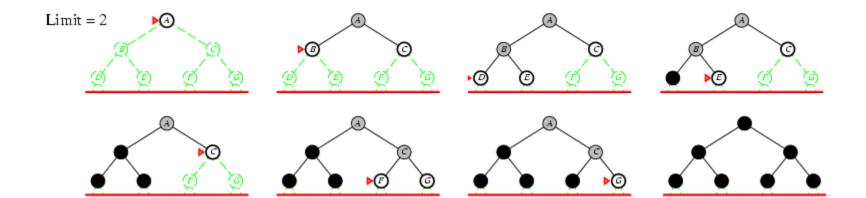




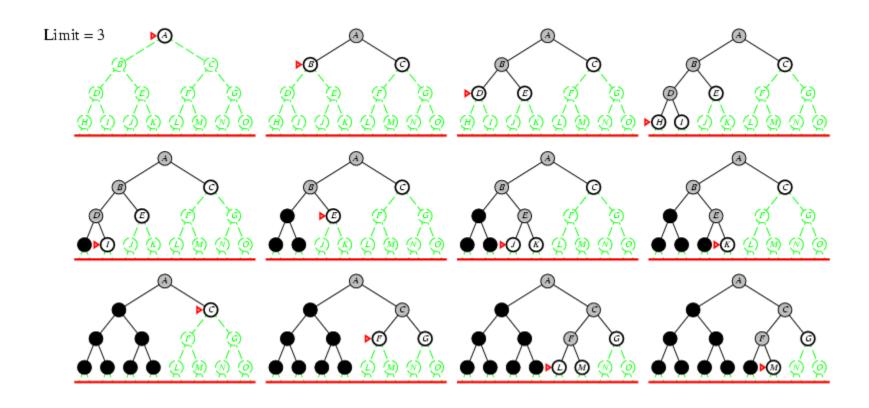




Iterative deepening search *IL*=2



Iterative deepening search *IL*=3



Iterative deepening search

• Number of nodes generated in a depth-limited search to depth *d* with branching factor *b*:

$$N_{DLS} = b^0 + b^1 + b^2 + \dots + b^{d-2} + b^{d-1} + b^d$$

• Number of nodes generated in an iterative deepening search to depth *d* with branching factor *b*:

$$N_{IDS} = (d+1)b^0 + db^1 + (d-1)b^2 + ... + 3b^{d-2} + 2b^{d-1} + 1b^d =$$

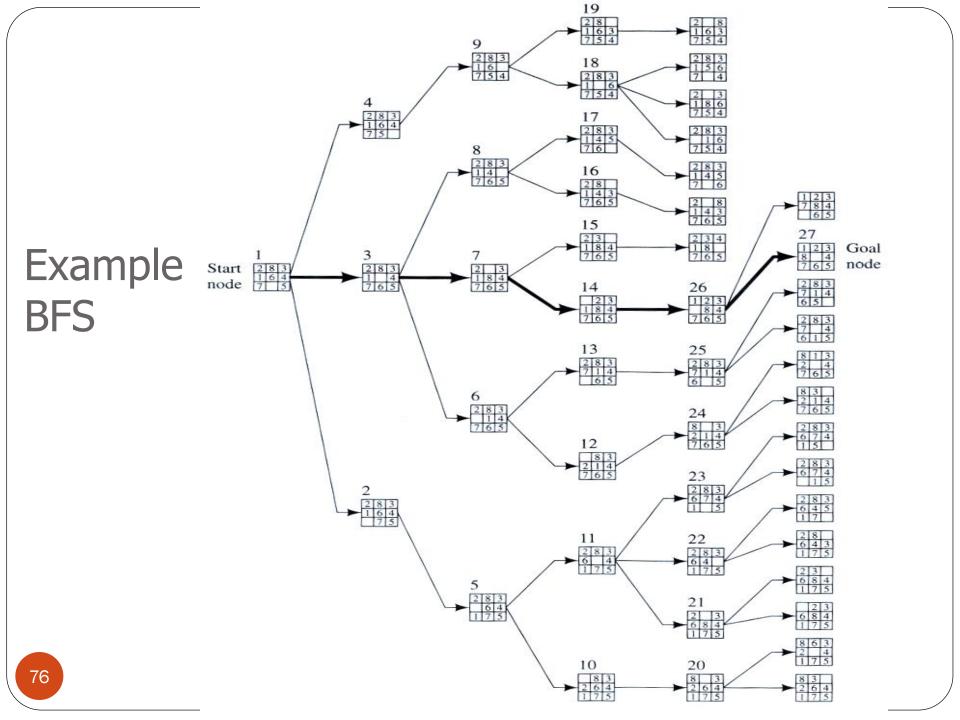
$$O(b^d) \neq O(b^{d+1})$$

BFS

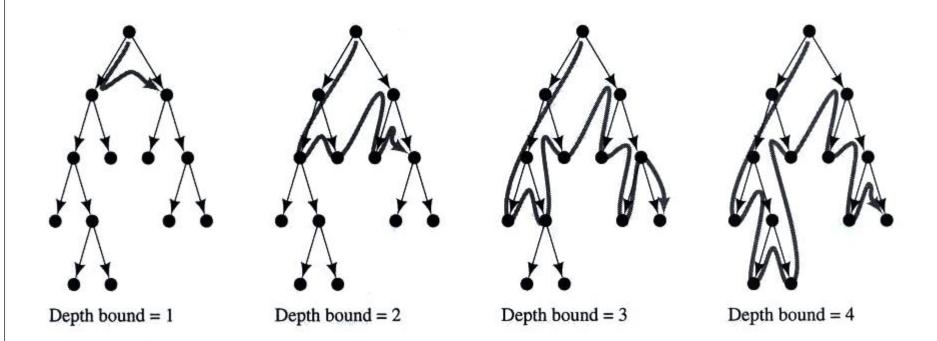
- For b = 10, d = 5,
 - $N_{DLS} = 1 + 10 + 100 + 1,000 + 10,000 + 100,000 = 111,111$
 - $N_{IDS} = 6 + 50 + 400 + 3,000 + 20,000 + 100,000 = 123,450$
 - $N_{BFS} = \dots = 1,111,100$

Properties of iterative deepening search

- <u>Complete?</u>Yes
- Time? $(d+1)b^0 + db^1 + (d-1)b^2 + \dots + b^d = O(b^d)$
- Space? O(bd)
- Optimal? Yes, if step cost = 1 or increasing function of depth.



Example IDS



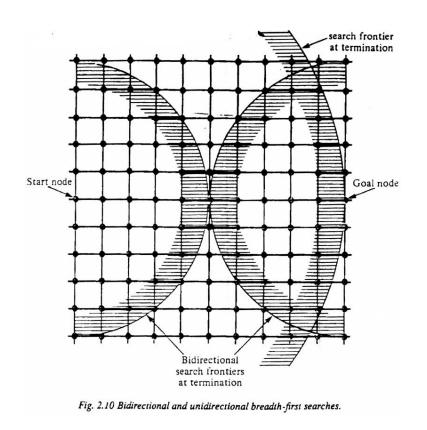
Stages in Iterative-Deepening Search

Bidirectional Search

- Idea
 - simultaneously search forward from S and backwards from G
 - stop when both "meet in the middle"
 - need to keep track of the intersection of 2 open sets of nodes
- What does searching backwards from G mean
 - need a way to specify the predecessors of G
 - this can be difficult,
 - e.g., predecessors of checkmate in chess?
 - what if there are multiple goal states?
 - what if there is only a goal test, no explicit list?

Bi-Directional Search

Complexity: time and space complexity are: $O(b^{d/2})$

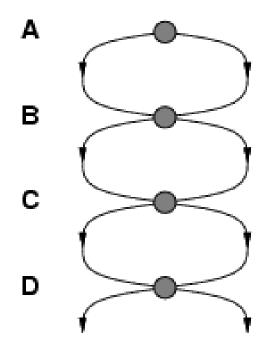


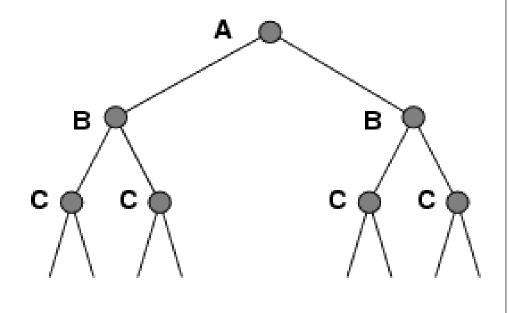
Summary of algorithms

Criterion	Breadth-	Uniform-	Depth-	Depth-	Iterative
	First	Cost	First	Limited	Deepening
Complete?	Yes	Yes	No	No	Yes
Time	$O(b^{d+1})$	$O(b^{\lceil C^*/\epsilon ceil})$	$O(b^m)$	$O(b^l)$	$O(b^d)$
Space	$O(b^{d+1})$	$O(b^{\lceil C^*/\epsilon ceil})$	O(bm)	O(bl)	O(bd)
Optimal?	Yes	Yes	No	No	Yes

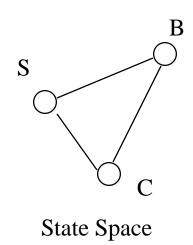
Repeated states

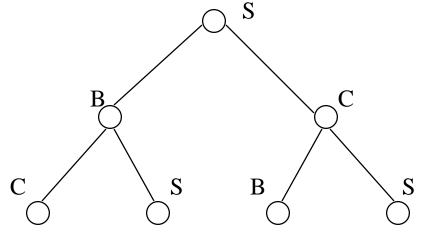
• Failure to detect repeated states can turn a linear problem into an exponential one!





Solutions to Repeated States





Example of a Search Tree

- Method 1suboptimal but practical
 - do not create paths containing cycles (loops)
- Method 2
 optimal but memory inefficient
 - never generate a state generated before
 - must keep track of all possible states (uses a lot of memory)
 - e.g., 8-puzzle problem, we have 9! = 362,880 states