

Lecture 4

Uninformed Search Strategies, Search with Costs (Ch 3.1-3.5, 3.7.3)

Announcements

- Assignment 1 out this week
- Use first edition of the textbook (2010), but feel free to check corresponding sections on the new edition

Today's Lecture

- ➡ • Recap from Previous lectures

- Depth first search
- Breadth first search
- Iterative deepening
- Search with costs
- Intro to heuristic search (time permitting)

Bogus Version of Generic Search Algorithm

Input: a graph a set of
start nodes

Boolean procedure $\text{goal}(n)$ that tests if n is a goal
node

$\text{frontier} := \{g : g \text{ is a goal node}\};$ While
frontier is not empty:

 select and remove path $\langle n_0, \dots, n_k \rangle$ from frontier;

 If $\text{goal}(n_k)$ return

$\langle n_0, \dots, n_k \rangle;$

 Find a neighbor n of n_k

 add n to frontier; end

- How many bugs?

A. One

B. Two

C.

Three D. Four

Bogus Version of Generic Search Algorithm

Input: a graph a set of
start nodes

Boolean procedure $\text{goal}(n)$ that tests if n is a goal
node

frontier := [$\langle g \rangle$: g is a goal node];

While frontier is not empty:

select and remove path $\langle n_o, \dots, n_k \rangle$ from frontier;

If $\text{goal}(n_k)$ return

$\langle n_o, \dots, n_k \rangle$;

Find a neighbor n of n_k

add n to frontier; end

- Start at the **start** node(s), NOT at the goal

- Add **all** neighbours of n_k to the frontier, NOT just one
- Add **path(s)** to frontier, NOT just the node(s)

Comparing Searching Algorithms:

Will it find a solution? the best one?

Def. : A search algorithm is **complete** if whenever there is at least one solution, the algorithm **is guaranteed to find it** within a finite amount of time.

Def.: A search algorithm is **optimal** if when it finds a solution, it is **the best one**

Def.: The **time complexity** of a search algorithm is the **worst- case** amount of time it will take to run, expressed in terms of

- maximum path length **m**
- maximum branching factor **b**.

Def.: The **space complexity** of a search algorithm is the **worst-case** amount of memory that the algorithm will use (i.e., the maximum number of nodes on the frontier), also expressed in terms of **m** and **b**.

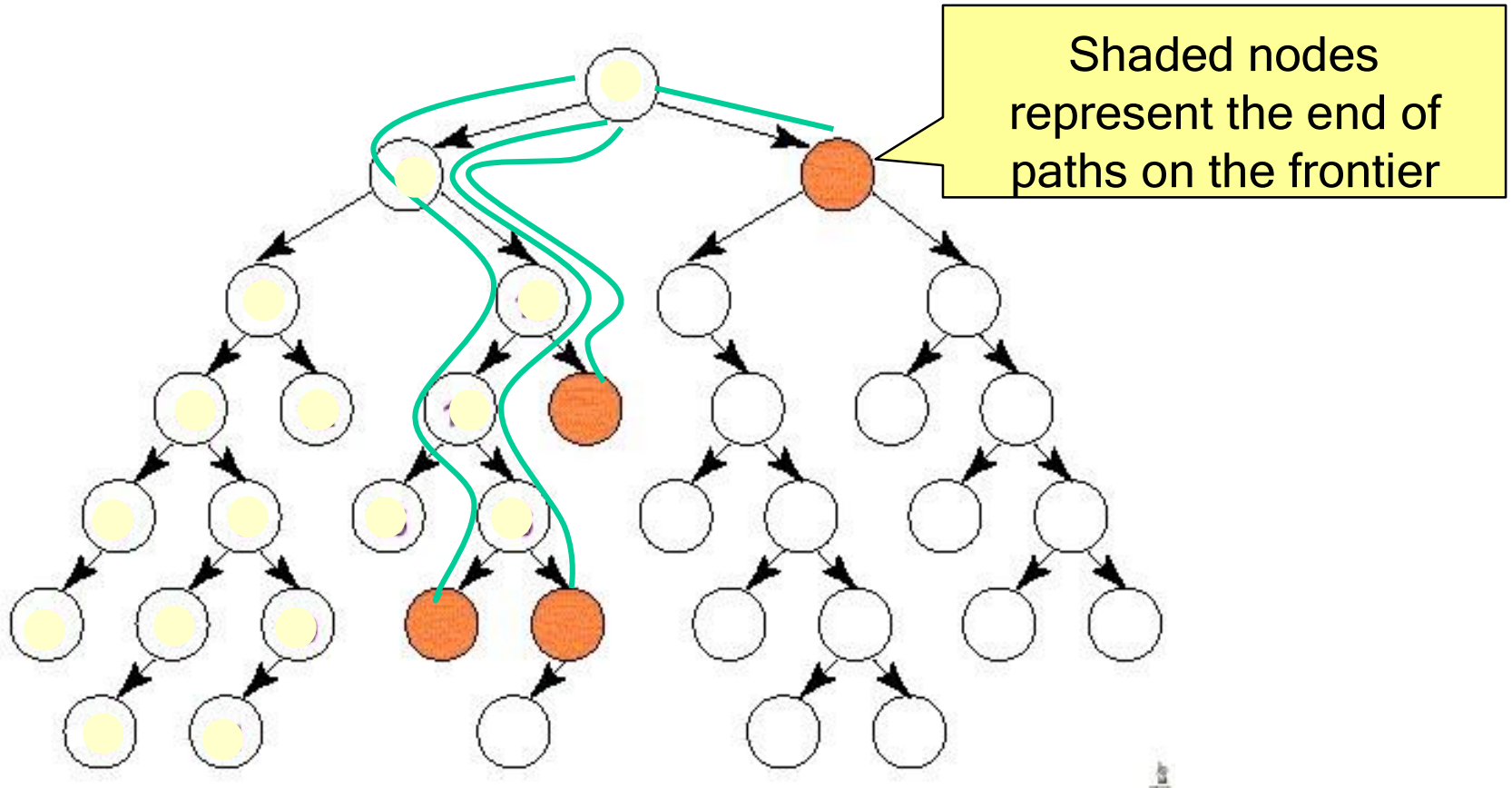
Slide 6

Today's Lecture

- Recap from Previous lectures
- ➡ • Depth first search
- Breadth first search
- Iterative deepening

- Search with costs
- Intro to heuristic search (time permitting)

Illustrative Graph: DFS



DFS explores each path on the frontier until its end (or until a goal is found) before considering any other path.

Input: a graph a set
of start
nodes
Boolean procedure $\text{goal}(n)$
testing if n is a goal node

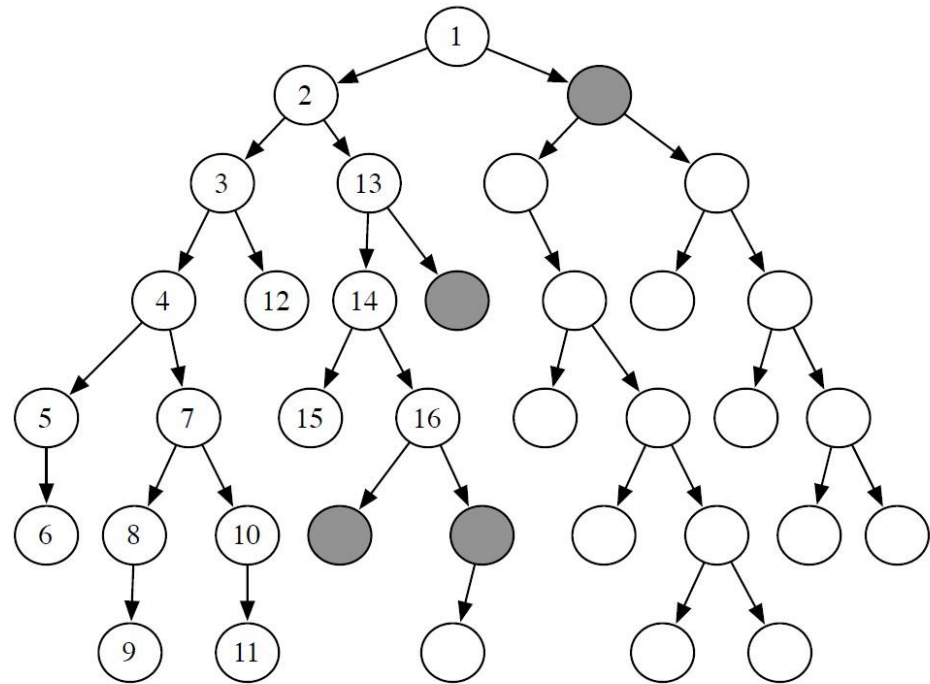
$\text{frontier} := [\langle s \rangle : s \text{ is a start}$
node]; While frontier is not
empty:

select and remove path $\langle n_0, \dots, n_k \rangle$
from frontier ;

If $\text{goal}(n_k)$

return

$\langle n_0, \dots, n_k \rangle$;



```
For every neighbor  $n$  of  $n_k$ ,  
  add  $\langle n_0, \dots, n_k, n \rangle$  to  
  frontier;
```

```
end
```

- In DFS, the frontier is a **last-in-first-out stack**

DFS as an instantiation of the Generic Search Algorithm

Let's see how this works in

DFS in AI Space

- Go to: <http://www.aispace.org/mainTools.shtml>
- Click on “Graph Searching” to get to the Search Applet
- Select the “Solve” tab in the applet
- Select one of the available examples via “File -> Load Sample Problem (good idea to start with the “Simple Tree” problem)
- Make sure that “Search Options -> Search Algorithms” in the toolbar is set to “Depth-First Search”.
- Step through the algorithm with the “Fine Step” or

“Step” buttons in the toolbar • The panel above the graph panel verbally describes what is happening during each step

- The panel at the bottom shows how the frontier evolves

See available help pages and video tutorials for more details on how to use the Search applet (<http://www.aispace.org/search/index.shtml>)

Slide 10

Depth-first Search: DFS

Example:

- the frontier is $[p_1, p_2, \dots, p_r]$ -each p_k is a path
- neighbors of last node of p_1 are $\{n_1, \dots, n_k\}$
- What happens?
- p_1 is selected, and its last node is tested for being a goal. If not

Analysis of DFS

- New paths are created by adding each of $\{n_1, \dots, n_k\}$ to p_1 • These new paths replace p_1 at the beginning of the frontier.
- Thus, the frontier is now: $[(p_1, n_1), \dots, (p_1, n_k), p_2, \dots, p_r]$.

NOTE: p_2 is only selected when all paths extending p_1 have been explored.

- You can get a much better sense of how DFS works by looking at the Search Applet in AI Space

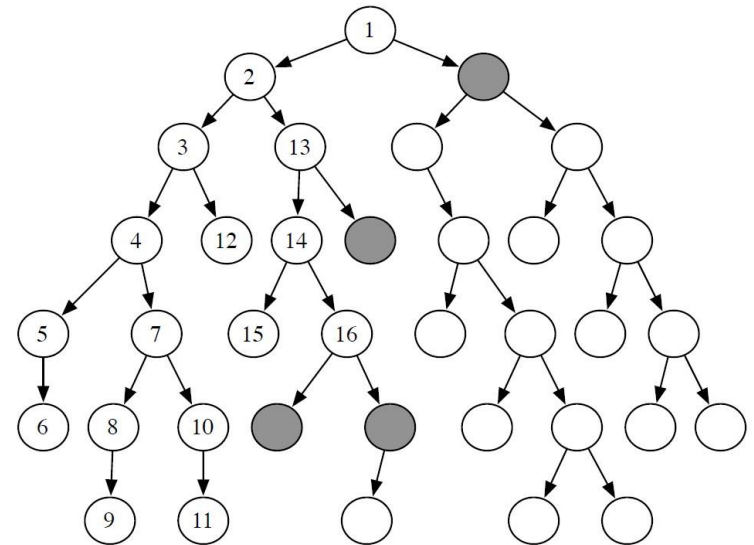
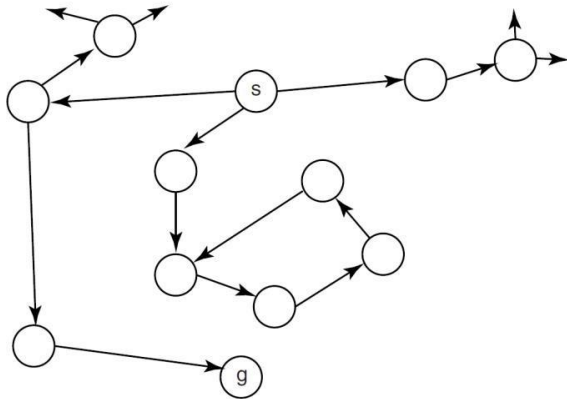
Slide 11

Def. : A search algorithm is **complete** if whenever there is at least one solution, the algorithm

is guaranteed to find it within a finite amount of time.

Is

DFS complete?

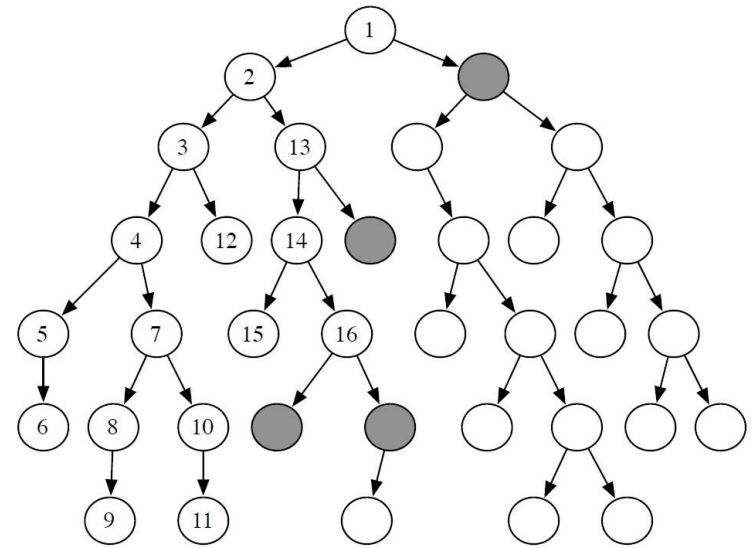
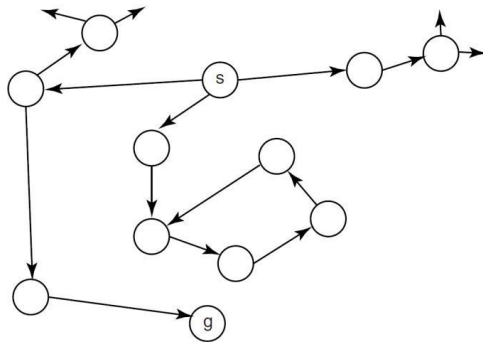


Analysis of DFS

Def. : A search algorithm is **complete** if whenever there is at least one solution, the algorithm **is guaranteed to find it** within a finite amount of time.

Is DFS complete?

No



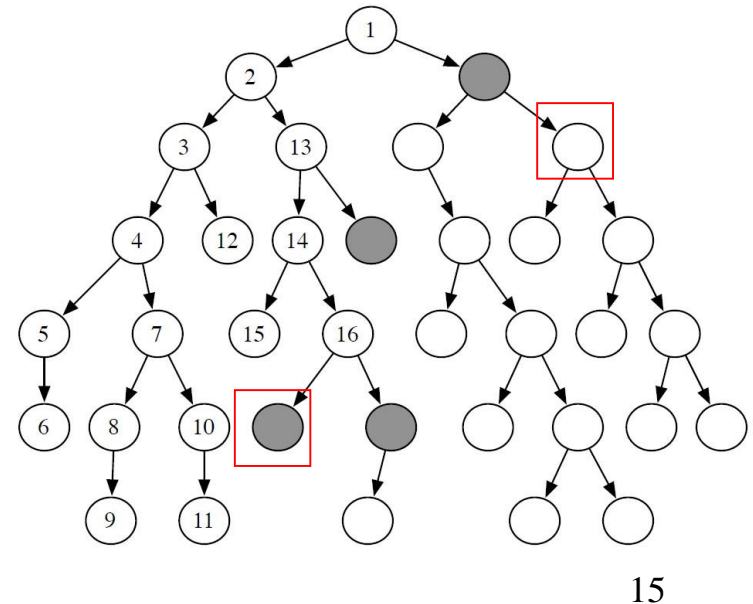
- If there are cycles in the graph, DFS may get “stuck” in one of them
- see this in AISpace by loading “Cyclic Graph Examples” or by adding a cycle to “Simple Tree”
- e.g., click on “Create” tab, create a new edge from N7 to N1, go back to ₁₃ “Solve” and see what happens

Def.: A search algorithm is **optimal** if when it finds a solution, it is **the best one** (e.g., the shortest)

Analysis of DFS

Is DFS **optimal**?

- E.g., goal nodes: red boxes



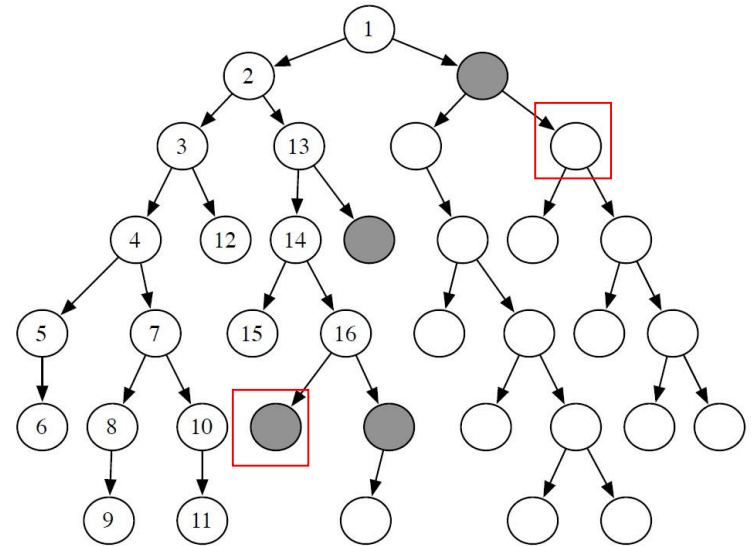
Def.: A search algorithm is **optimal** if when it finds a solution, it is **the best one** (e.g., the shortest)

Is DFS **optimal**?

A Yes

B No

- E.g., goal nodes: red boxes

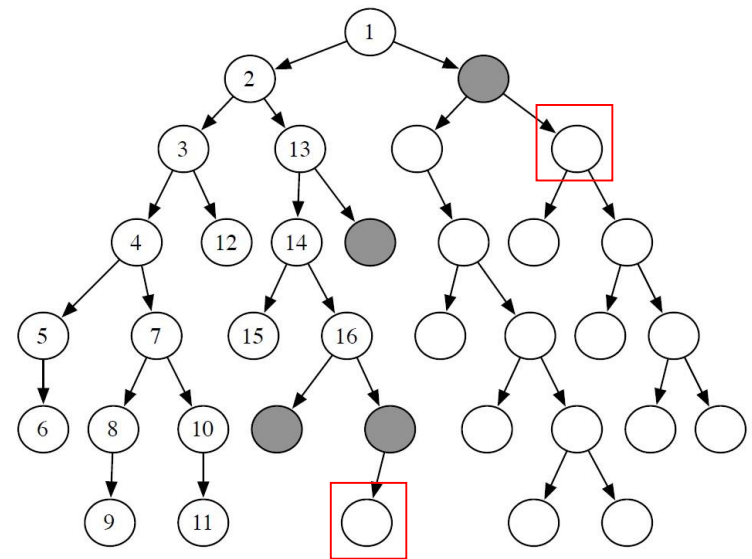


Analysis of DFS

Def.: A search algorithm is **optimal** if when it finds a solution, it is **the best one** (e.g., the shortest)

Is DFS **optimal**? **No**

- It can “stumble” on longer solution paths before it gets to shorter ones.
- E.g., goal nodes: red boxes
- see this in AISpace by loading “Extended Tree Graph” and set N6 as a goal
- e.g., click on “Create” tab, right-click on N6 and select “set as a goal node”



Analysis of DFS

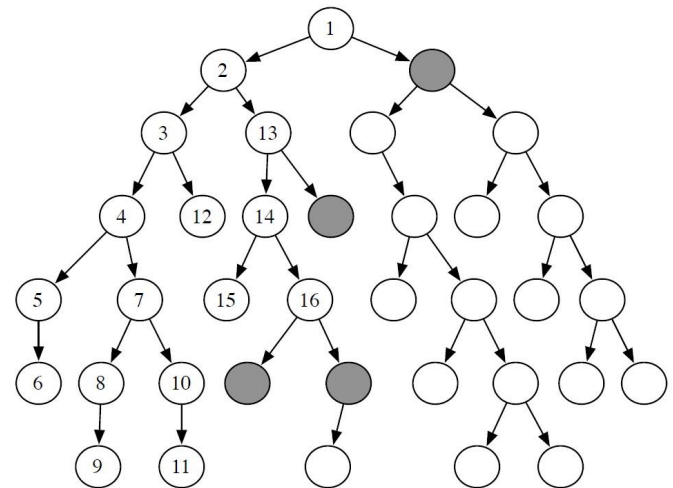
Def.: The **space complexity** of a search algorithm is the **worst-case** amount of memory that the algorithm will use (i.e., the maximal number of nodes on the frontier), expressed in terms of

- maximum path length m
- maximum forward branching factor b .

Analysis of DFS

- What is DFS's **space complexity**, in terms of **m** and **b** ?

See how
this works
in



Analysis of DFS

Def.: The **space complexity** of a search algorithm is the **worst-case** amount of memory that the algorithm will use (i.e., the maximal number of nodes on the frontier), expressed in terms of

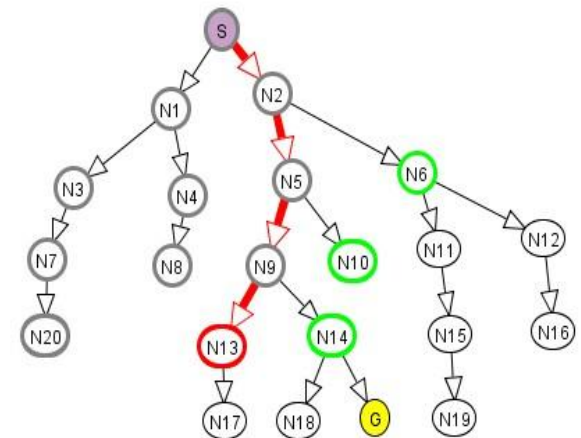
- maximum path length m
- maximum forward branching factor b .

- What is DFS's **space complexity**, in terms of m and b ?

Analysis of DFS

$O(bm)$

- for every node in the path currently explored, DFS maintains a path to its unexplored siblings in the search tree
- Alternative paths that DFS needs to explore
- The longest possible path is m , with a maximum of $b-1$ alternative paths per node along the path



Algorithm Selected: Depth First

CURRENT PATH:

S → N2 → N5 → N9 → N13

NEW FRONTIER:

Node: N14 Path Cost: 33.3

Node: N10 Path Cost: 21.7

Node: N6 Path Cost: 23.4

Path: S → N2 → N5 → N9 → N14

Path: S → N2 → N5 → N10

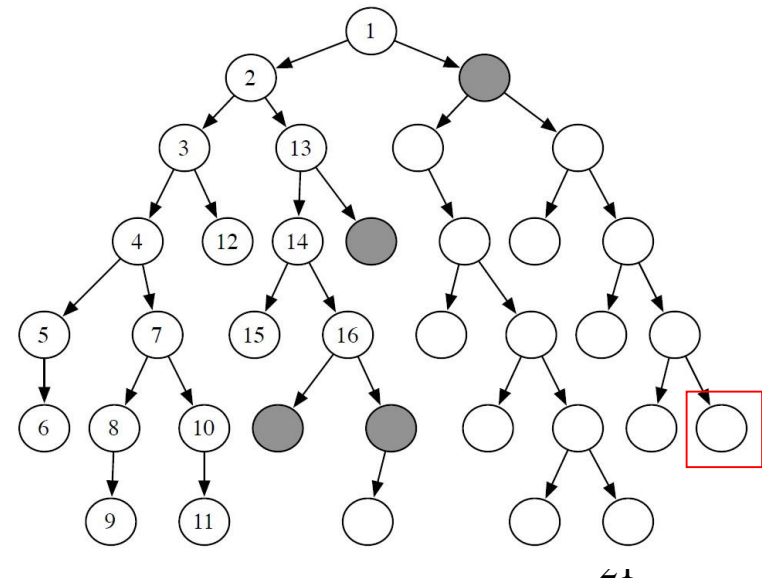
Path: S → N2 → N6

Analysis of DFS

Def.: The **time complexity** of a search algorithm is the **worst-case** amount of time it will take to run, expressed in terms of

- maximum path length m
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- What is DFS's **time complexity**, in terms of **m** and **b**?



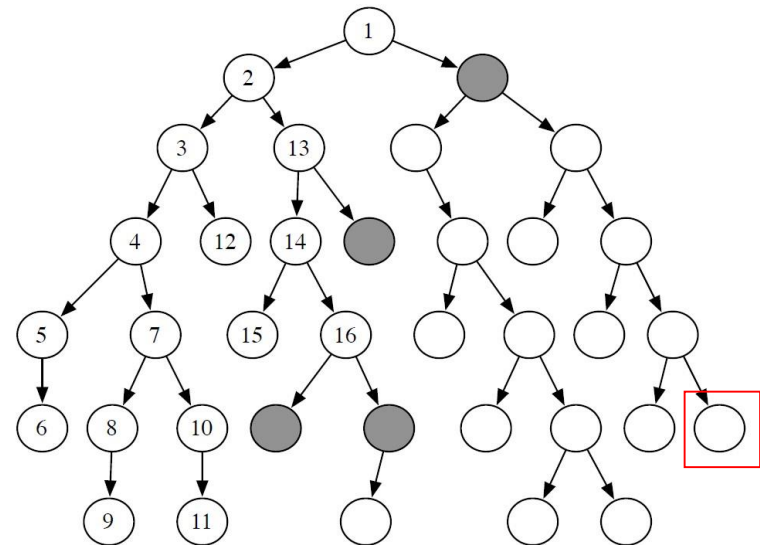
- E.g., single goal node -> red box

Analysis of DFS

Def.: The **time complexity** of a search algorithm is the **worst-case** amount of time it will take to run, expressed in terms of

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- What is DFS's **time complexity**, in terms of **m** and **b**?



A.
 $O(b^m)$

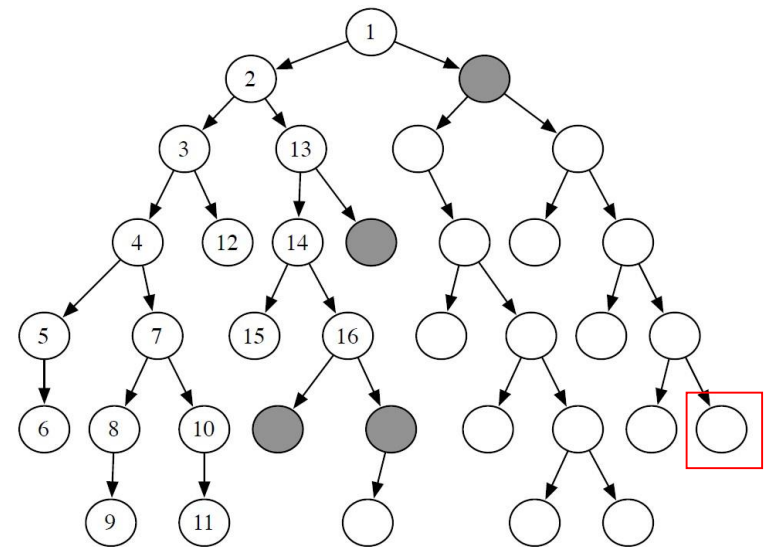
B.
 $O(m^b)$

C.
 $O(bm)$

D.
 $O(b+m)$

- E.g., single goal node -> red box
- Hint: think about how many nodes are in a search tree m steps away from the start

Analysis of DFS



Def.: The **time complexity** of a search algorithm is the **worst-case** amount of time it will take to run, expressed in terms of

- maximum path length m
- maximum forward branching factor b .

- What is DFS's **time complexity**, in terms of m and b ?

$$O(b^m)$$

- In the worst case, must examine every node in the tree
- E.g., single goal node -> red box

To Summarize

	Complete	Optimal	Time	Space
DFS	NO	NO	$O(b^m)$	$O(bm)$



Analysis of DFS: Summary

- Is DFS **complete**? NO
- Depth-first search isn't guaranteed to halt on graphs with cycles.
- However, DFS is complete for **finite acyclic graphs**.
- Is DFS **optimal**? NO
- It can “stumble” on longer solution paths before it gets to shorter ones.
- What is the worst-case **time complexity**, if the maximum path length is m and the maximum branching factor is b ?

- $O(b^m)$: must examine every node in the tree.
- Search is unconstrained by the goal until it happens to stumble on the goal.
- What is the worst-case space complexity?
- $O(bm)$
- the longest possible path is m , and for every node in that path must maintain a fringe of size b . Slide

Analysis of DFS (cont.)

DFS is appropriate when

- Space is restricted
- Many solutions, with long path length

It is a poor method when

- There are cycles in the graph
- There are sparse solutions at shallow depth

Why DFS need to be studied and understood?

- It is simple enough to allow you to learn the basic aspects of searching

- It is the basis for a number of more sophisticated and useful search algorithms (e.g. Iterative Deepening)

Today's Lecture

- Recap from Previous lectures
- Depth first search - analysis
- ➡ • Breadth first search
- Iterative deepening
- Search with costs
- Intro to heuristic search (time permitting)

- BFS explores all paths of length l on the frontier,

before looking at path of length $l + 1$

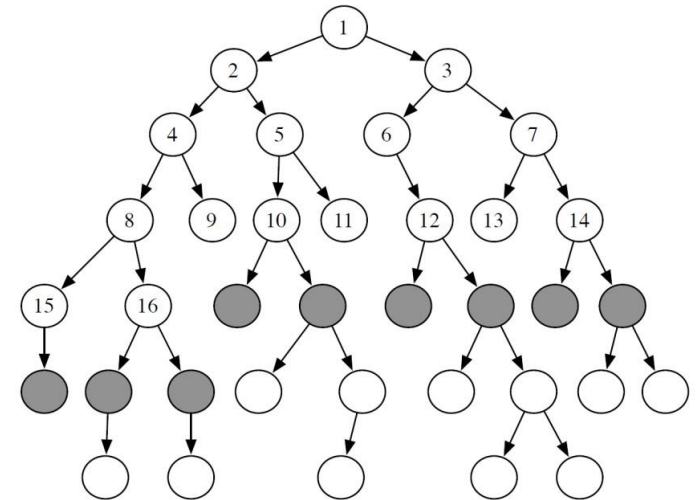


```
Input: a graph a set of start
      nodes
      Boolean procedure goal(n)
      testing if n is a goal node
frontier:= [<s>: s is a start node]; While
frontier is not empty:
    select and remove path <n0,...,nk> from frontier;
    If goal(nk) return
      <n0,...,nk>;
    Else
      For every neighbor n of nk, add
        <n0,...,nk, n> to frontier;
end
```

Let's see how this works in AIspace

in the Search Applet toolbar, set “Search Options -> Search Algorithms” to “Breadth-First Search”.

BFS as an instantiation of the Generic Search Algorithm



In BFS, the frontier is a
first-in-first-out queue

Breadth-first Search: BFS

Example:

- the frontier is $[p_1, p_2, \dots, p_r]$
- neighbors of the last node of p_1 are $\{n_1, \dots, n_k\}$
- What happens?
- p_1 is selected, and its end node is tested for being a goal. If not
- New k paths are created attaching each of $\{n_1, \dots, n_k\}$ to p_1
- These follow p_r at the end of the frontier.

- Thus, the frontier is now $[p_2, \dots, p_r, (p_1, n_1), \dots, (p_1, n_k)]$.
- p_2 is selected next.

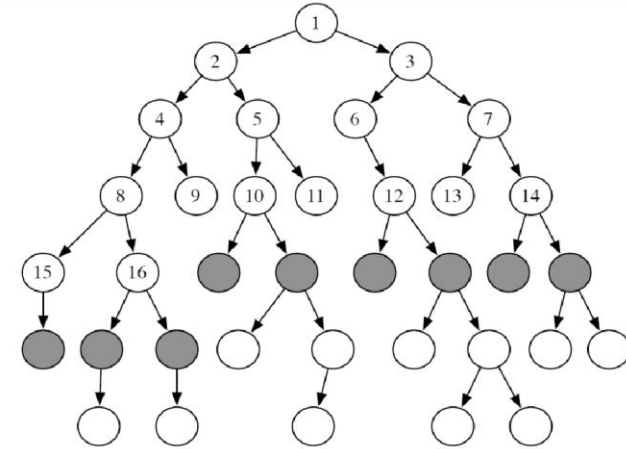
As for DFS, you can get a much better sense of how BFS works by looking at the Search Applet in AI Space



Slide

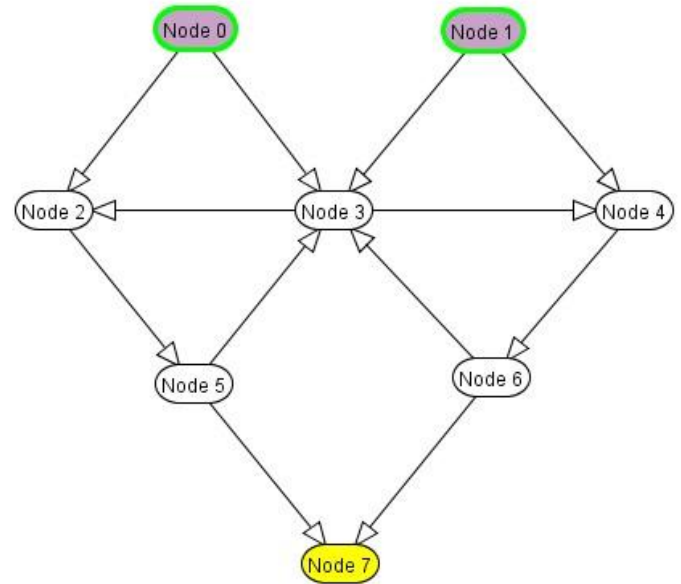
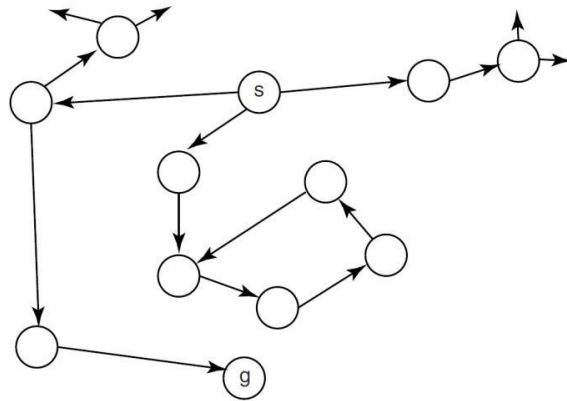
Analysis of BFS

Def. : A search algorithm is **complete** if whenever there is at least one solution, the algorithm is **guaranteed to find it** within a finite amount of time.



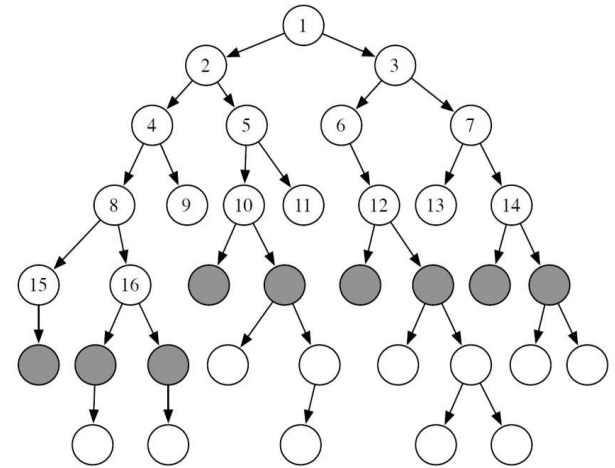
Is BFS **complete**?

Analysis of BFS



Analysis of BFS

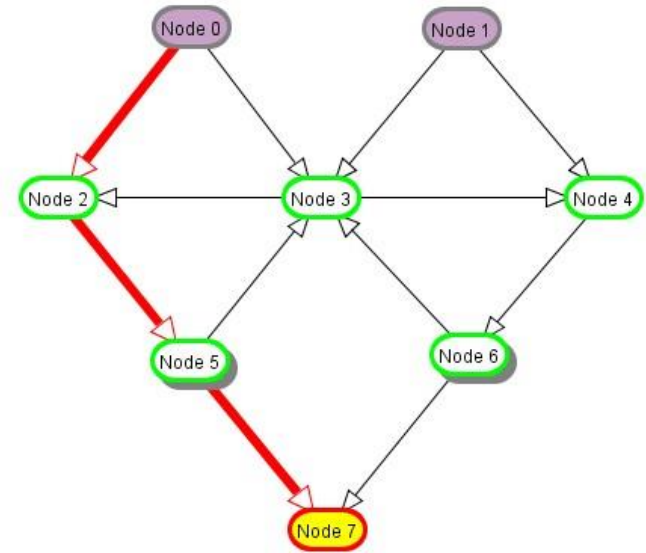
Def. : A search algorithm is **complete** if whenever there is at least one solution, the algorithm **is guaranteed to find it** within a finite amount of time.



Analysis of BFS

Is BFS complete? **Yes**

- If a solution exists at level l , the path to it will be explored before any other path of length $l + 1$
- impossible to fall into an infinite cycle
- see this in AISpace by loading “Cyclic Graph Examples” or by adding a cycle to “Simple Tree”



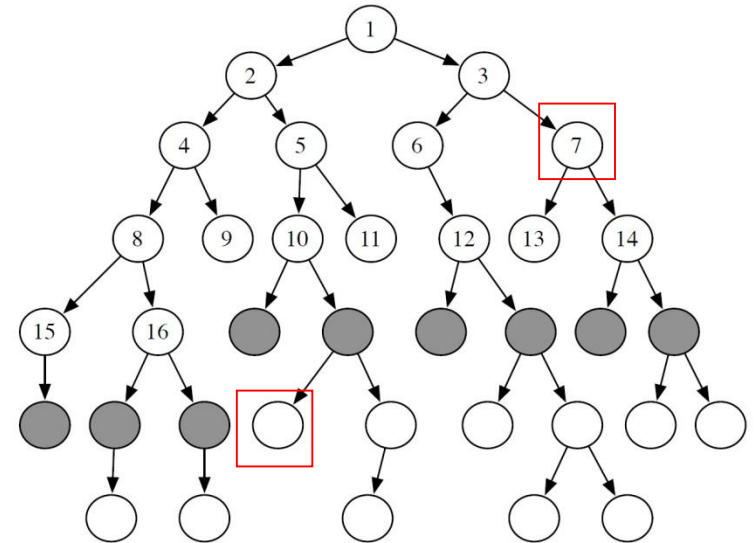
Def.: A search algorithm is **optimal** if when it finds a solution, it is **the best one**

Analysis of BFS

Is BFS **optimal**?

- E.g., two goal nodes: red boxes

Def.: A search algorithm is **optimal** if when it finds a solution, it is **the best one**

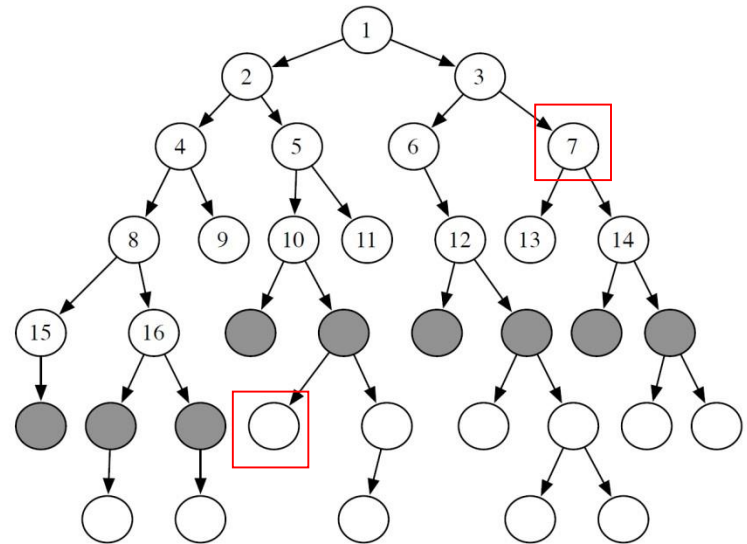


Analysis of BFS

Is BFS optimal?

Yes

- E.g., two goal nodes: red boxes
- Any goal at level l (e.g. red box N7) will be reached before goals at lower levels



Analysis of BFS

Def.: The **time complexity** of a search algorithm is the **worst-case** amount of time it will take to run, expressed in terms of

- maximum path length m
- maximum forward branching factor b .

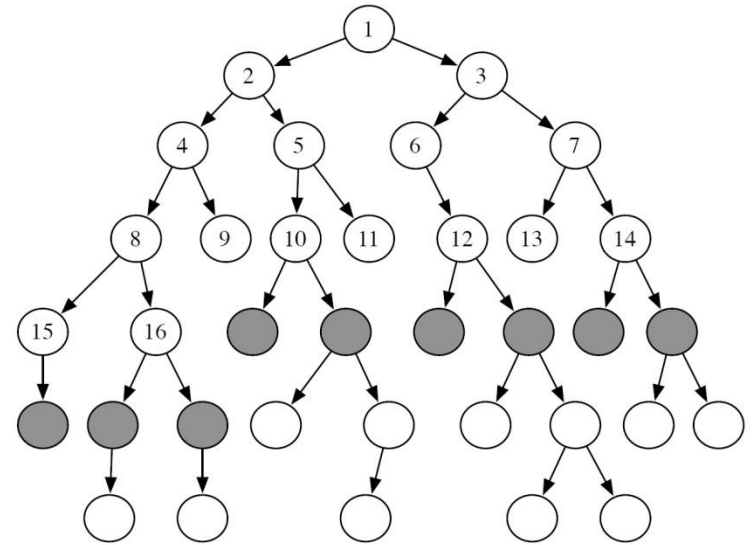
- What is BFS's **time complexity**, in terms of m and b ?

A. $O(b^m)$

B. $O(m^b)$

Analysis of BFS

C. $O(bm)$



D. $O(b+m)$

Def.: The **time complexity of** a search algorithm is the **worst-case** amount of time it will take to run, expressed in terms of

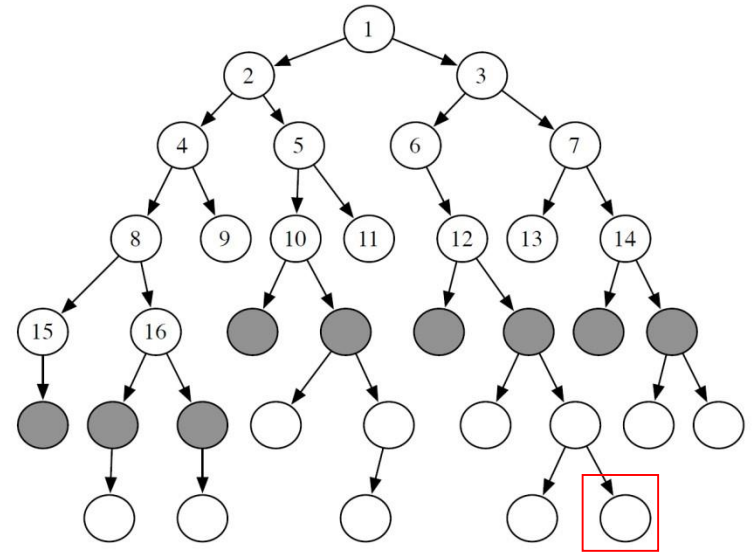
- maximum path length m
- maximum forward branching factor b .

$O(b^m)$

- What is BFS's **time complexity**, in terms of m and b ?

Analysis of BFS

- Like DFS, in the worst case BFS must examine every node in the tree
- E.g., single goal node -> red box

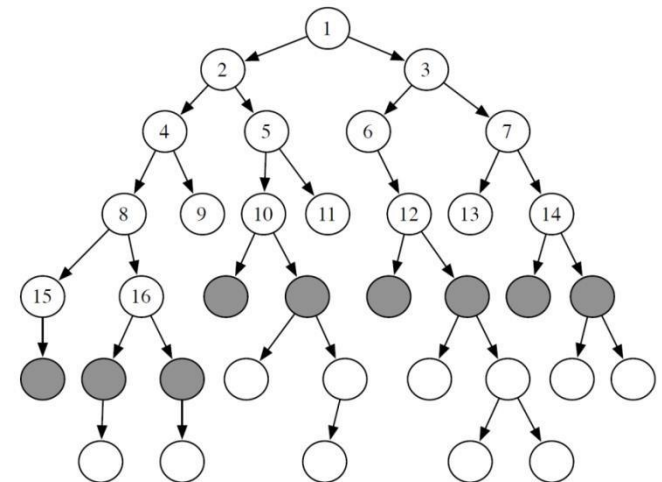


Analysis of BFS

Def.: The **space complexity** of a search algorithm is the **worst case** amount of memory that the algorithm will use (i.e., the maximal number of nodes on the frontier), expressed in terms of

- maximum path length m
- maximum forward branching factor b .

- What is BFS's **space complexity**, in terms of m and b ?



Analysis of BFS

Def.: The **space complexity** of a search algorithm is the **worst case** amount of memory that the algorithm will use (i.e., the maximal number of nodes on the frontier), expressed in terms of

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A. $O(b^m)$

B. $O(m^b)$

C. $O(bm)$

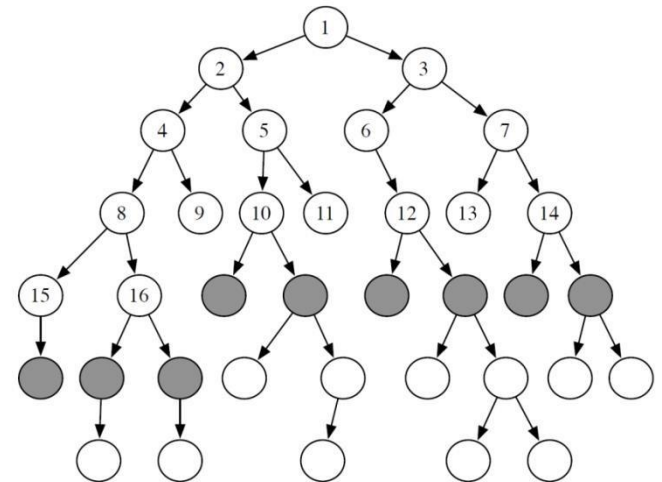
D. $O(b+m)$

Analysis of BFS

- What is BFS's **space complexity**, in terms of **m** and **b** ?

$$O(b^m)$$

- BFS must keep paths to all the nodes at level m



When to use BFS vs. DFS?

- The search graph has cycles or is infinite
- We need the shortest path to a solution
- There are only solutions at great depth
- There are some solutions at shallow depth
- No way the search graph will fit into memory

When to use BFS vs. DFS?

- The search graph has cycles or is infinite

BFS

- We need the shortest path to a solution

BFS

- There are only solutions at great depth

DFS

- There are some solutions at shallow depth

BFS

- No way the search graph will fit into memory

DFS

To Summarize

	Complete	Optimal	Time	Space
DFS	NO	NO	$O(b^m)$	$O(bm)$
BFS	YES	YES	$O(b^m)$	$O(b^m)$

How can we achieve an acceptable (linear) space complexity while maintaining completeness and optimality?

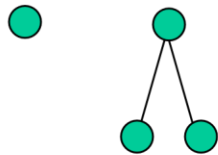
Key Idea: re-compute elements of the frontier rather than saving them.



Iterative Deepening DFS (IDS) in a Nutshell

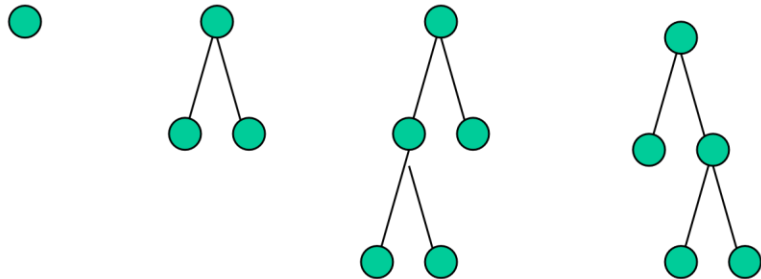
- Depth-bounded depth-first search

depth = 1



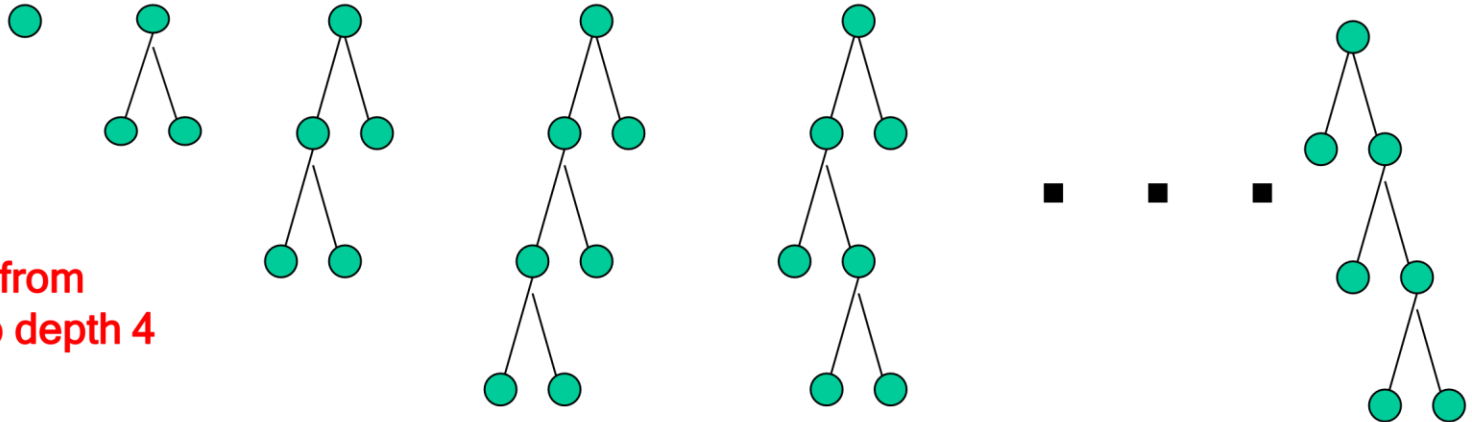
If no goal re-start from scratch and get to depth 2

depth = 2



If no goal re-start from scratch and get to depth 3

depth = 3



If no goal re-start from scratch and get to depth 4

(Time) Complexity of IDS

- Use DFS to look for solutions at depth 1, then 2, then 3, etc
 - For depth D , ignore any paths with longer length

(Time) Complexity of IDS

- That sounds wasteful!
- Let's analyze the time complexity
- For a solution at depth m with branching factor b

(Time) Complexity of IDS

Depth	Total # of paths at that level	#times created by BFS (or DFS)	#times created by IDS	Total #paths for IDS
1				
2				
.
.
.
m-1				
m				

- That sounds wasteful!

(Time) Complexity of IDS

- Let's analyze the time complexity
- For a solution at depth m with branching factor b

Depth	Total # of paths at that level	#times created by BFS (or DFS)	#times created by IDS	Total #paths for IDS
1	b	1	m	mb
2				
.				
.				
.				
$m-1$				
m				

(Time) Complexity of IDS

- That sounds wasteful!
- Let's analyze the time complexity
- For a solution at depth m with branching factor b

Depth	Total # of paths at that level	#times created by BFS (or DFS)	#times created by IDS	Total #paths for IDS
1	b	1	m	mb
2	b^2	1	$m-1$	$(m-1) b^2$
.				
.				
.				
$m-1$				
m				

(Time) Complexity of IDS

- That sounds wasteful!
- Let's analyze the time complexity
- For a solution at depth m with branching factor b

Depth	Total # of paths at that level	#times created by BFS (or DFS)	#times created by IDS	Total #paths for IDS
1	b	1	m	mb
2	b^2	1	$m-1$	$(m-1) b^2$
.
.
.
$m-1$	b_{m-1}	1	2	$2 b_{m-1}$
m	b_m	1	1	b_m

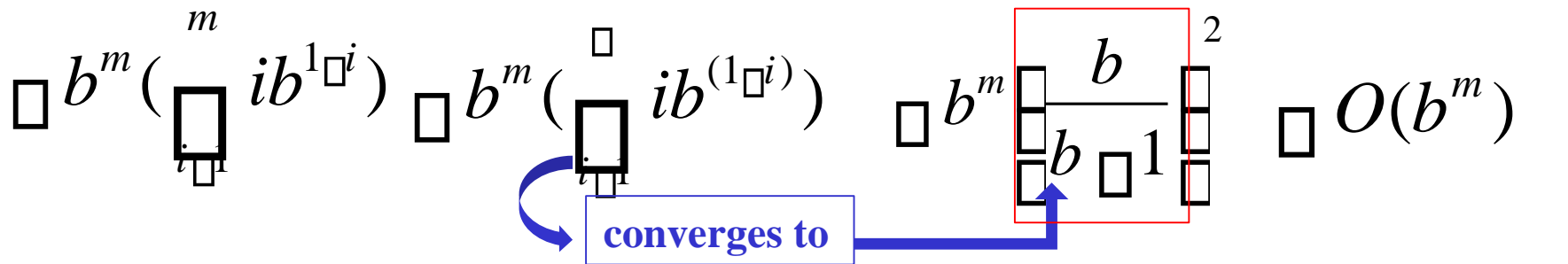
(Time) Complexity of IDS

Solution at depth m , branching factor b

Total # of paths generated:

$$b^m + 2 b^{m-1} + 3 b^{m-2} + \dots + mb$$

$$= b^m (1 b^0 + 2 b^{-1} + 3 b^{-2} + \dots + m b^{1-m})$$



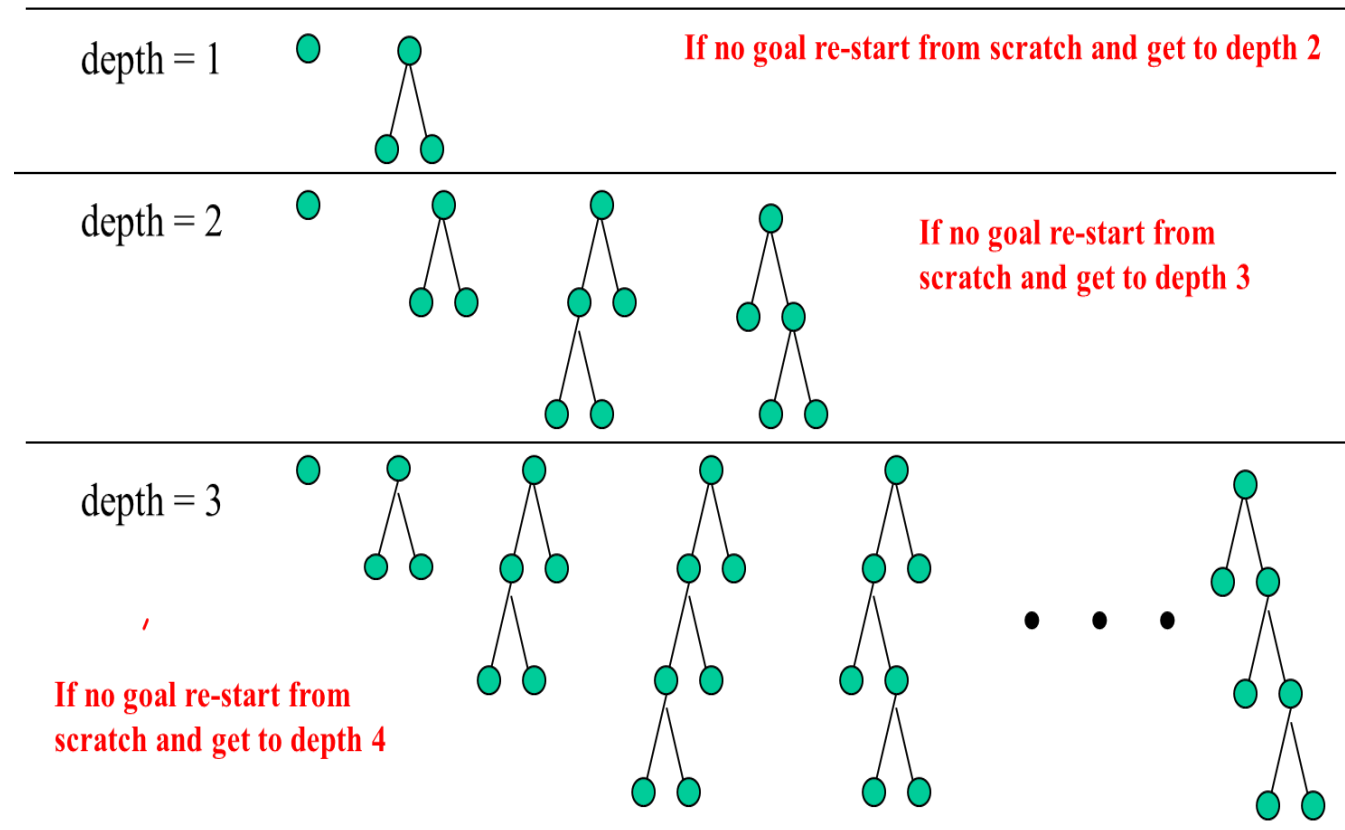
- For $b=10$, $m=5$. BSF 111,111 and ID = 123,456 (only 11% more nodes)

(Time) Complexity of IDS

- The larger b the better, but even with $b = 2$ the search ID will take only 2 times as much as BFS

Further Analysis of Iterative Deepening DFS (IDS)

- Space complexity $O(mb)$
- Same as DFS!



- Complete?

- Optimal?

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Further Analysis of Iterative Deepening DFS (IDS)

- Space complexity

$O(mb)$

- DFS scheme, only explore one branch at a time
- Complete?

Yes

- Only paths up to depth m , doesn't explore longer paths
 - cannot get trapped in infinite cycles, gets to a solution first

- Optimal? Yes

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Summary of Uninformed Search

	Complete	Optimal	Time	Space
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DFS	N	N	$O(b^m)$	$O(mb)$
BFS	Y	Y (shortest)	$O(b^m)$	$O(b^m)$
IDS	Y	Y (shortest)	$O(b^m)$	$O(mb)$
LCFS				

Learning Goals for today's class

- **Select** the most appropriate search algorithms for specific problems.
- Depth-First Search vs. Breadth-First Search vs. Iterative Deepening
- **Define/read/write/trace/debug** different search algorithms

TO DO

- Heuristic Search and A*: Ch 3.6, 3.6.1

