



Posts and Telecommunication Institute of Technology
Faculty of Information Technology 1

Introduction to Artificial Intelligence

Solving problems by searching

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Outline

- ▶ The search problem in state spaces
- ▶ Examples
- ▶ Basic search algorithms

Outline

- ▶ The search problem in state spaces
 - Search and AI
 - Search problem formulation
 - Criteria for evaluating search algorithms
- ▶ Examples
- ▶ Basic search algorithms

Search & AI

- ▶ Many problems can be formulated as a search problem
 - Game: find an **optimal movement** (take advantages)
 - Planning: find a **solution that satisfies requirements** (constraint satisfaction)
 - Route finding: find the **optimal path** (length, time, cost, ...)

- ▶ Search is an important research direction of AI
 - Developing efficient search algorithms (especially in cases where the search space is large)
 - Foundations of many other research branches of AI
 - Machine learning, Natural language processing, inference

Search problem formulation

A search problem can be formulated through 5 components

1. A finite set of possible **states**: Q
2. A set of **initial states**: $S \subseteq Q$
3. **Operator** or **successor function** $P(x)$, set of states reachable from x by any single action
4. **Goal test**:
 - Explicitly: a set of possible goal states $G \subseteq Q$
 - Implicitly: specified by an abstract property
5. **Path cost**
 - Sum of the costs of the individual actions along the path
 - $c(x, a, y) \geq 0$, cost moving from state x to state y by taking action a

A **solution** is a path (a series of actions) from the initial state to a goal state

Criteria for evaluating search algorithms

- ▶ Computational complexity (time complexity)
 - The amount of computation required to find the solution
 - **Number of states to consider** before finding a solution
- ▶ Space complexity
 - **Number of states to store concurrently** in memory when executing the algorithm
- ▶ Completeness
 - Is the algorithm **guaranteed to find a solution** when there is one?
- ▶ Optimality
 - Does the algorithm **find the highest-quality solution** when there are several different solutions?

Outline

- ▶ The search problem in state spaces
- ▶ Examples
 - 8-puzzle game
 - 8-queen problem
- ▶ Basic search algorithms

8-puzzle game (1 / 2)

7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

Goal State

(Russell & Norvig, 2010)

8-puzzle game (2/2)

7	2	4
5		6
8	3	1

Start State

	1	2
3	4	5
6	7	8

Goal State

(Russell & Norvig, 2010)

- ▶ **States**: combination of cell positions
- ▶ **Initial state**: arbitrary state
- ▶ **Action**: move the empty cell up, down, left, right
- ▶ **Goal**: a pre-defined goal state
- ▶ **Cost**: number of movements

8-queen problem (1 / 2)

Place 8 queens on an 8x8 chess board so that no queens threaten each other

8-queen problem (2/2)

Place 8 queens on an 8x8 chess board so that no queens threaten each other

- ▶ **State**: arrangement of 0 to 8 queens on the board
- ▶ **Initial state**: no queen on the board
- ▶ **Action**: place a queen on an empty cell
- ▶ **Goal**: 8 queens on an 8x8 chess board so that no queens threaten each other

Outline

- ▶ The search problem in state spaces
- ▶ Examples
- ▶ **Basic search algorithms**
 - General search algorithm
 - Breadth-first search (BFS)
 - Uniform-cost search (UCS)
 - Depth-first search (DFS)
 - Iterative deepening search (IDS)

General search algorithm (1 / 3)

- ▶ **General idea:** consider states, using successor functions to extend those states until the desired state is reached

- ▶ Expanding states creates a “search tree”
 - Each state is a node
 - **Open nodes** are nodes waiting for further expansion
 - Expanded nodes are called **closed nodes**

General search algorithm (2/3)

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)

while($O \neq \emptyset$) **do**

1. Select a node $n \in O$ and delete n from O
2. **if** $n \in G$, **return** (path to n)
3. Add $P(n)$ to O

return: no solution

General search algorithm (3/3)

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)

while($O \neq \emptyset$) **do**

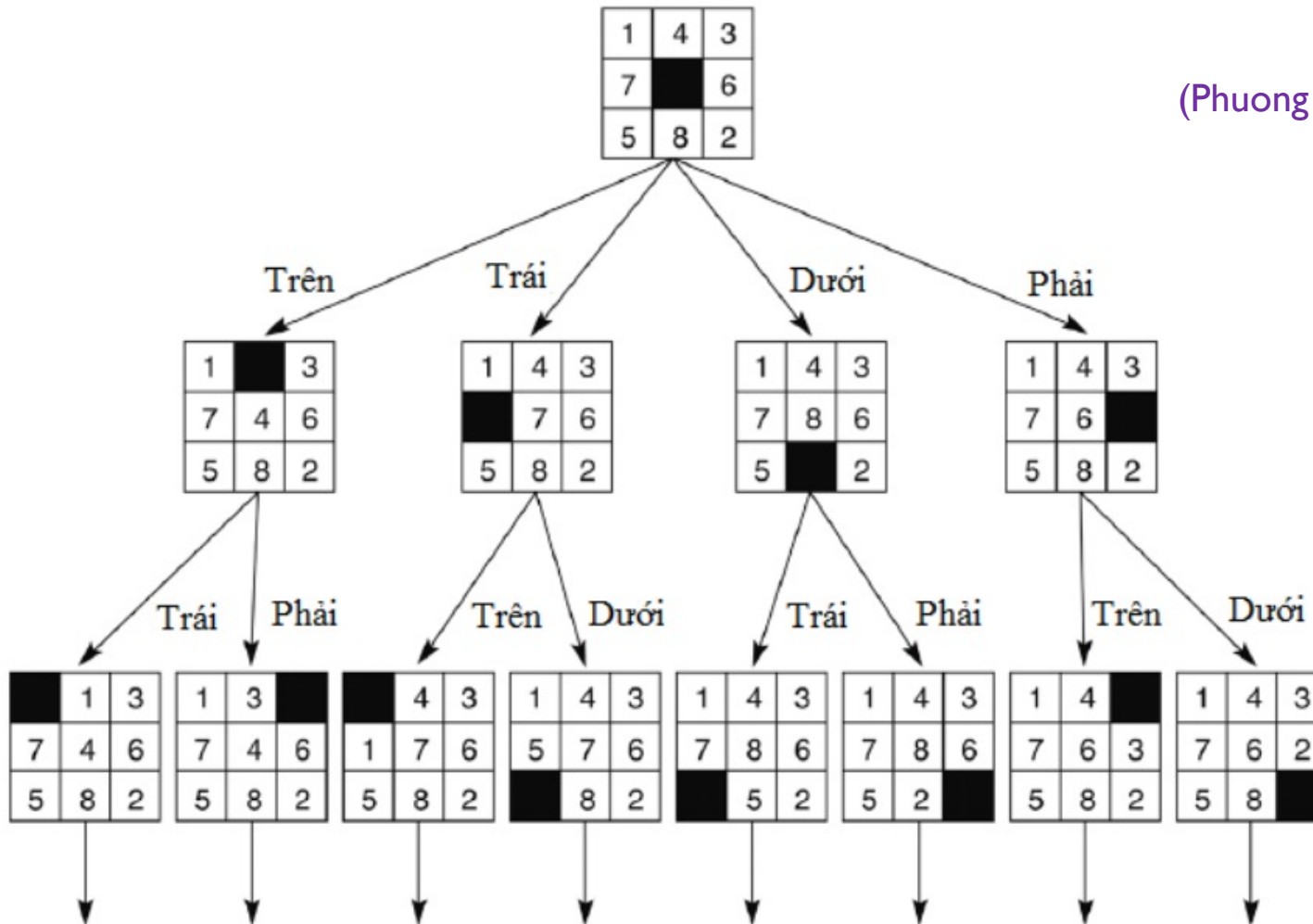
1. Select a node $n \in O$ and delete n from O
2. **if** $n \in G$, **return** (path to n)
3. Add $P(n)$ to O

return: no solution

How to select
node n ?

Example of a search tree

(Phuong TM, 2016)



Search strategies

- ▶ A search strategy is determined by **the order** in which the **nodes** in the search tree **are expanded**

- ▶ Criteria for evaluating search strategies:
 - **Completeness**: is guaranteed to find a solution (when there is)?
 - **Computational complexity**: number of generated nodes
 - **Space complexity**: number of nodes stored concurrently in memory
 - **Optimality**: does the algorithm find the best solution?
- ▶ The complexity is calculated based on the following parameters
 - b : branching factor (of the search tree)
 - d : depth of the solution
 - m : maximum depth of the state space (may be ∞)

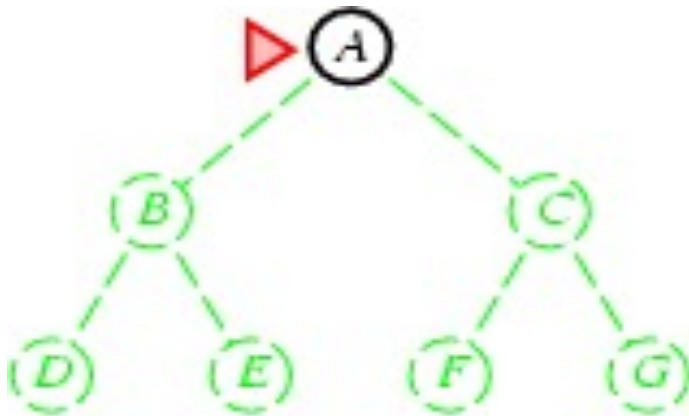
Blind search (Uninformed search)

- ▶ Blind search only uses information according to the problem statement during the search process

- ▶ Blind search algorithms
 - Breadth-first search (BFS)
 - Uniform-cost search (UCS)
 - Depth-first search (DFS)
 - Iterative deepening search (IDS)

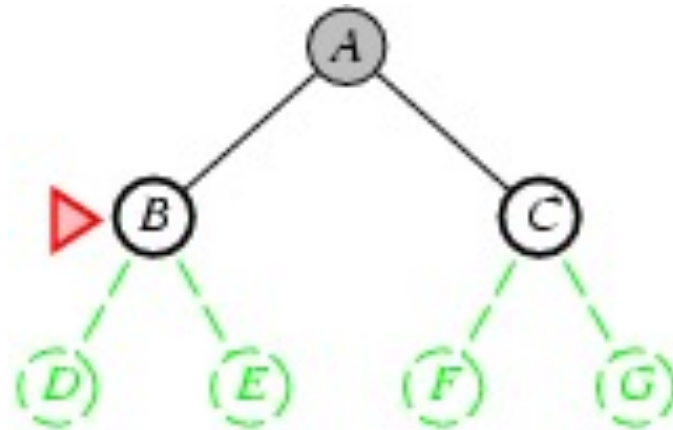
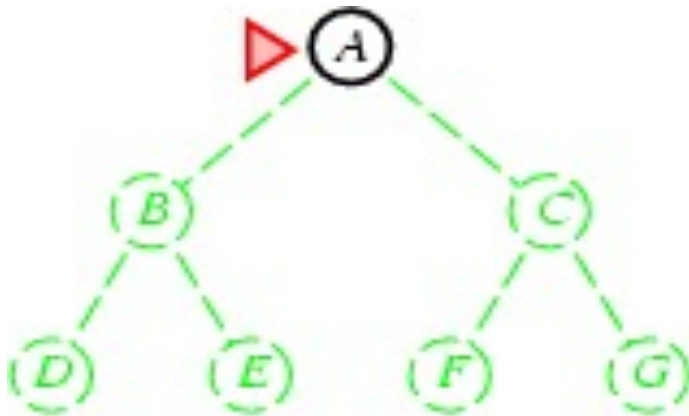
Breadth-first search – BFS (1/4)

- ▶ **Principle:** among open nodes, choose the shallowest node (closest to the root) to expand



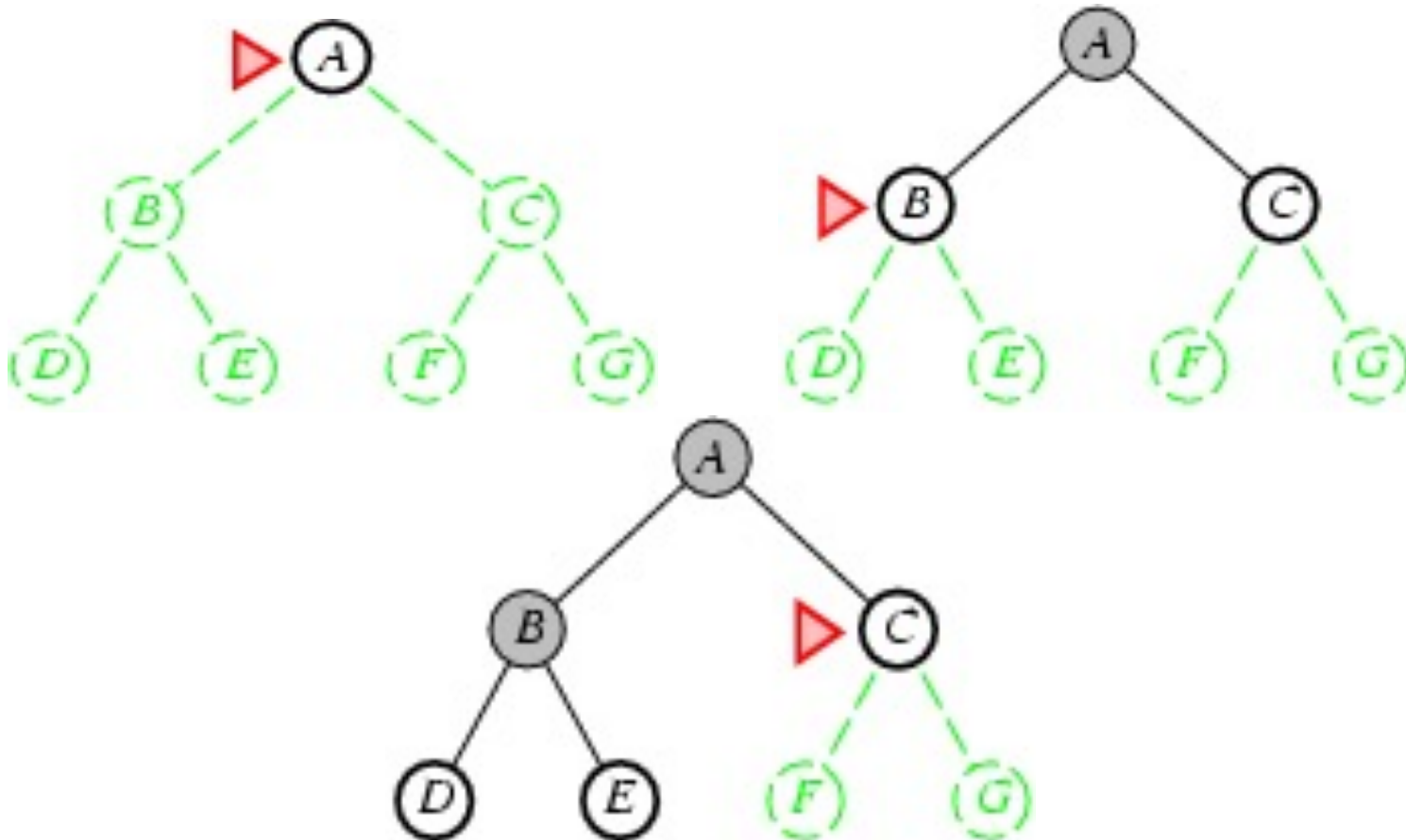
Breadth-first search – BFS (2/4)

- ▶ **Principle:** among open nodes, choose the shallowest node (closest to the root) to expand



Breadth-first search – BFS (3/4)

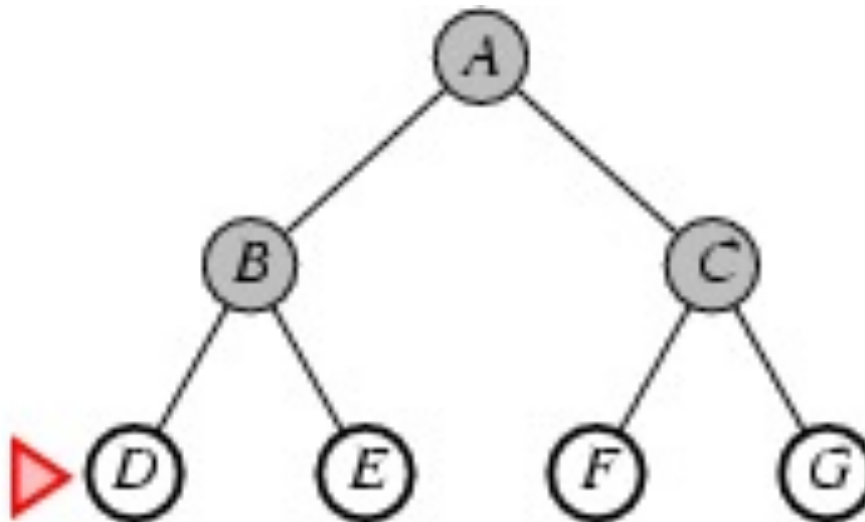
- ▶ **Principle:** among open nodes, choose the shallowest node (closest to the root) to expand



Breadth-first search – BFS (4/4)

► Remember the path

- When switching to a node, remember the parent node of that node by using a back pointer
- After reaching the goal, the back pointer is used to find the path back to the initial node



BFS algorithm (1 / 2)

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)

while($O \neq \emptyset$) **do**

1. take the **first node** n from O
2. **if** $n \in G$, **return** (path to n)
3. add $P(n)$ **to the end** of O

return no solution

BFS algorithm (2/2)

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)

while($O \neq \emptyset$) **do**

1. take the **first node** n from O
2. **if** $n \in G$, **return** (path to n)
3. add $P(n)$ **to the end** of O

return no solution

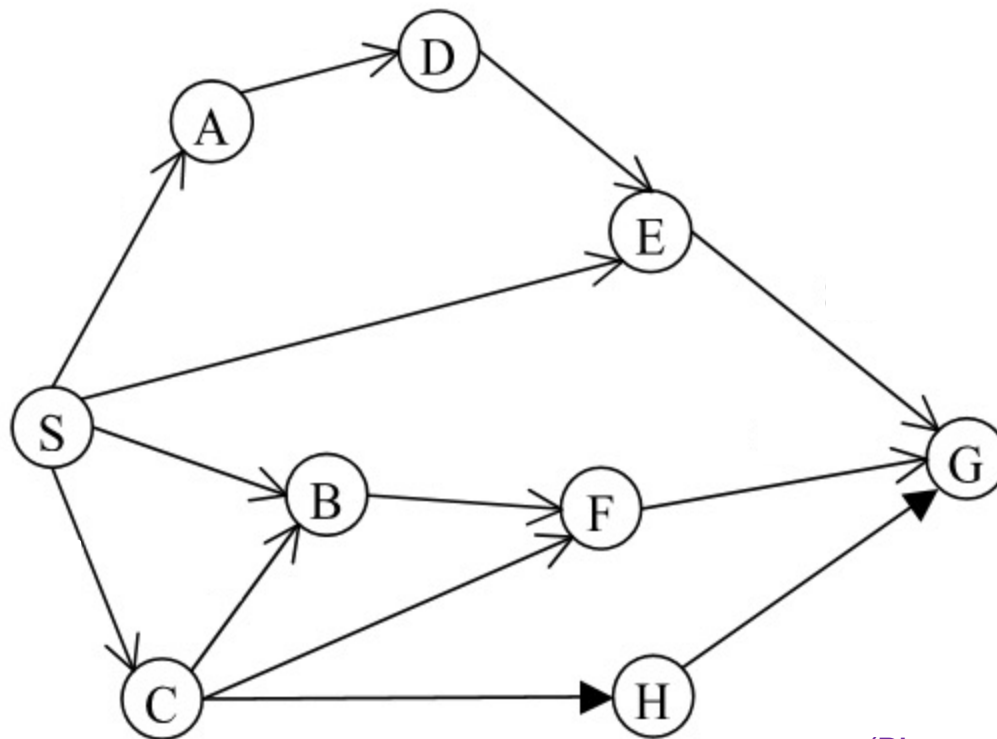
FIFO data
structure
(Queue)

Avoiding repeated nodes

- ▶ There can be **multiple paths** reaching to a node
 - The algorithm can **expand a node many times**
 - Can lead to an infinite loop

- ▶ **Solution**
 - Do not add a node to the queue if the node is already expanded or is in the queue (waiting to be expanded)
 - Remember less nodes; check goal faster
 - Avoid the infinite loop

BFS example (1 / 2)



(Phuong TM, 2016)

BFS example (2/2)

#	Expanded node	Open node list O (Queue)
0		S
1	S	A_S, B_S, C_S, E_S
2	A_S	B_S, C_S, E_S, D_A
3	B_S	C_S, E_S, D_A, F_B
4	C_S	E_S, D_A, F_B, H_C
5	E_S	D_A, F_B, H_C, G_E
6	D_A	F_B, H_C, G_E
7	F_B	H_C, G_E
8	H_C	G_E
9	G_E	Goal

Path: $G \leftarrow E \leftarrow S$

Properties of BFS

- ▶ **Completeness?**
 - Yes (if b is finite)
- ▶ **Time?**
 - $1 + b + b^2 + b^3 + \dots + b^d = O(b^d)$
- ▶ **Space?**
 - $O(b^d)$ (store all nodes)
- ▶ **Optimality?**
 - Yes (if cost = 1 for every action)
 - Always search all the nodes at the higher level before searching nodes at the lower level

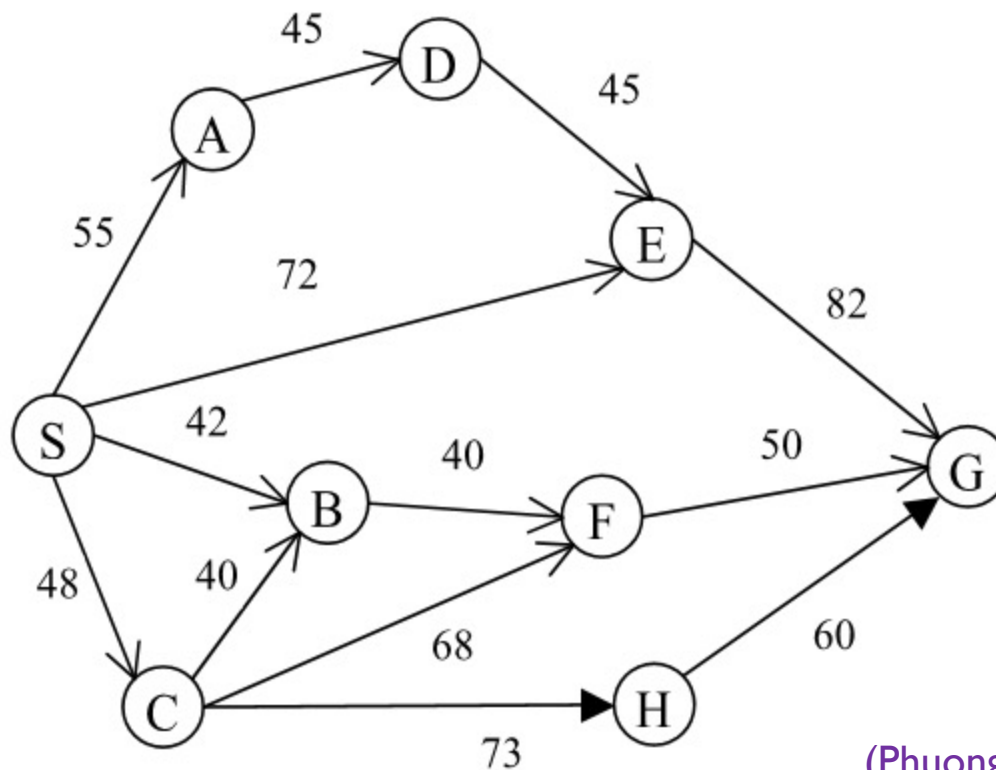
- ▶ **Memory** is more important than time

Uniform-cost search (UCS)

- ▶ When the costs of actions (moving between two nodes) are different
 - BFS does not give an optimal solution
 - Need to use uniform-cost search (a variant of BFS)

- ▶ **Principle:** choose the node with the **smallest cost** to expand first instead of choosing the shallowest node like in BFS

UCS example (1 / 2)



(Phuong TM, 2016)

UCS example (2/2)

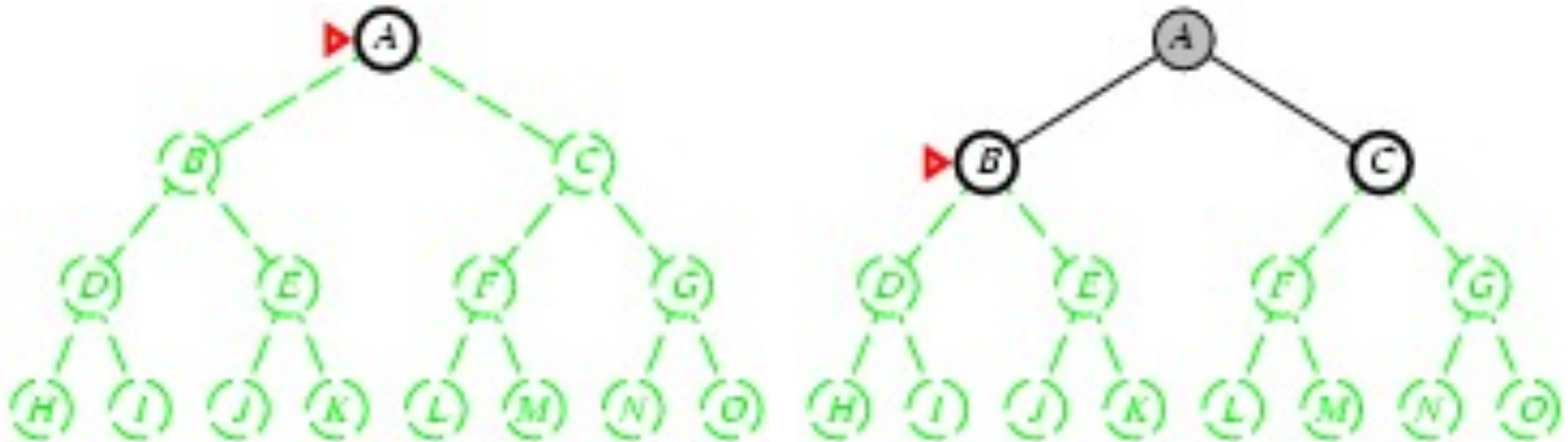
#	Expanded node	Open node list O
0		$S(0)$
1	S	$A_S(55), B_S(42), C_S(48), E_S(72)$
2	B_S	$A_S(55), C_S(48), E_S(72), F_B(82)$
3	C_S	$A_S(55), E_S(72), F_B(82), H_C(121)$
4	A_S	$E_S(72), F_B(82), H_C(121), D_A(100)$
5	E_S	$F_B(82), H_C(121), D_A(100), G_E(154)$
6	F_B	$H_C(121), D_A(100), G_F(132)$
7	D_A	$H_C(121), G_F(132)$
8	H_C	$G_F(132)$
9	G_F	Goal

Update
path to G

Path: $G \leftarrow F \leftarrow B \leftarrow S$

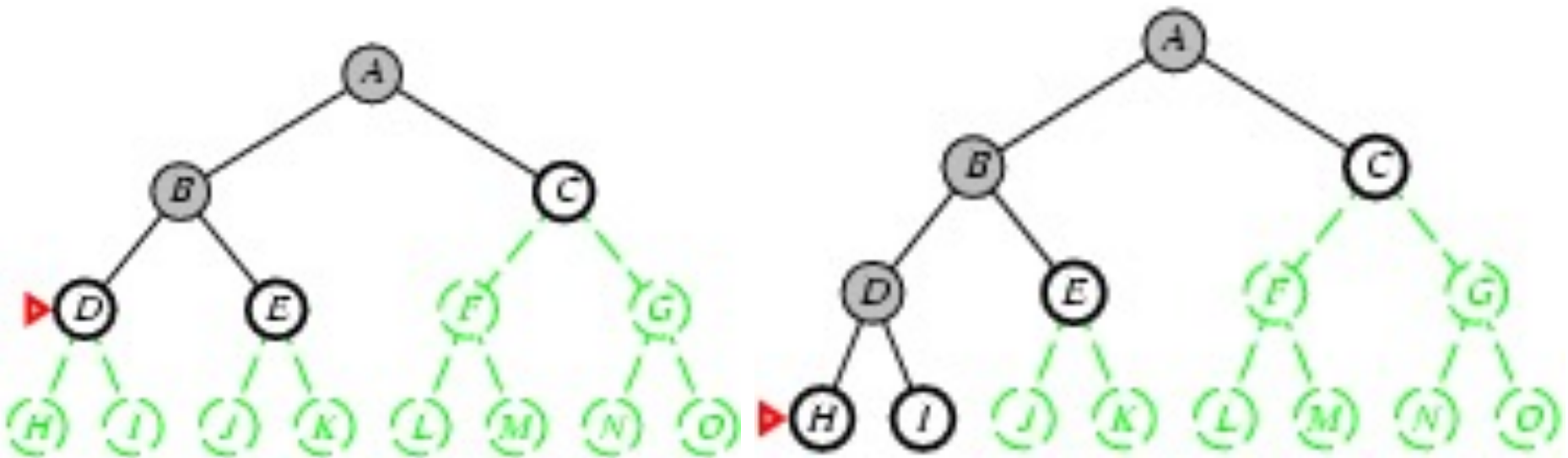
Depth-first search - DFS (1/4)

- ▶ **Principle:** among open nodes, choose the deepest node (farthest to the root) to expand



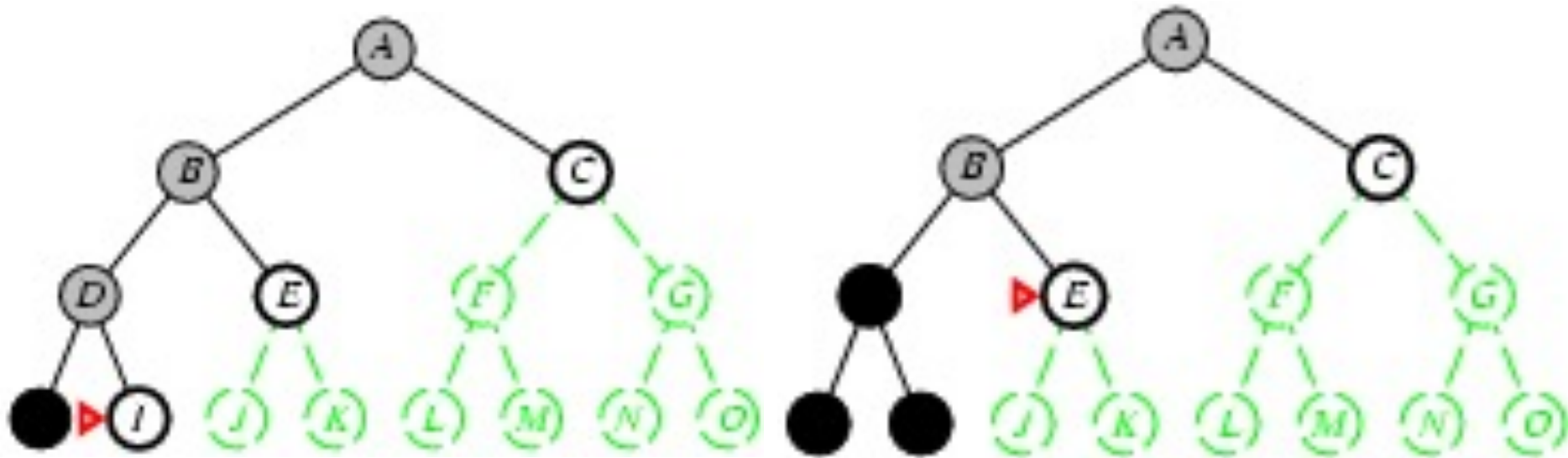
Depth-first search - DFS (2/4)

- ▶ **Principle:** among open nodes, choose the deepest node (farthest to the root) to expand



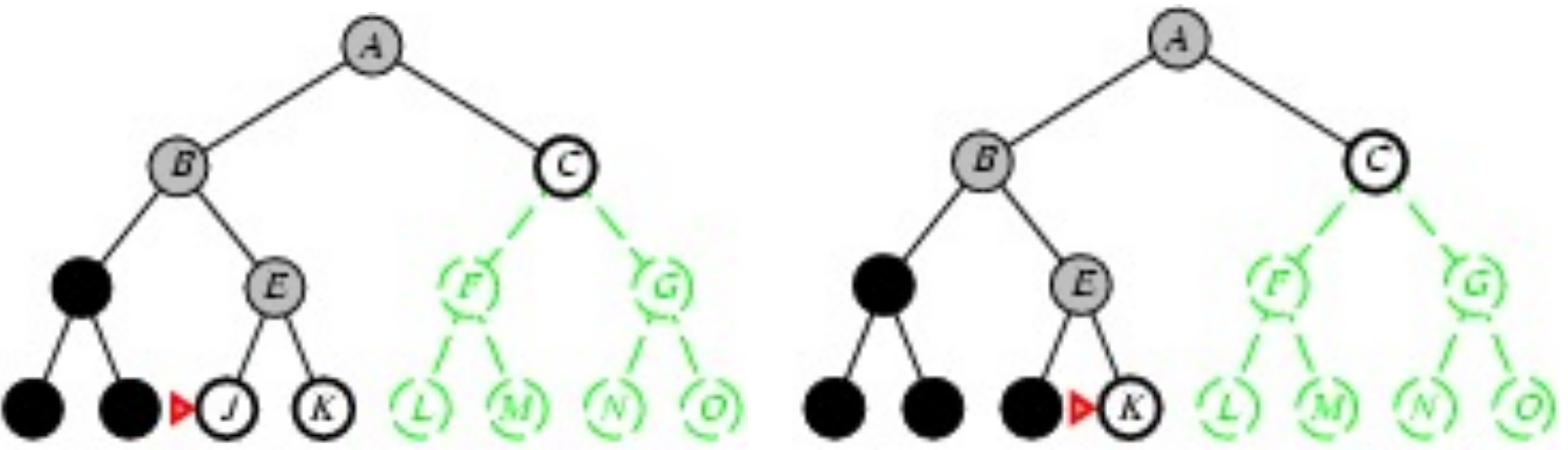
Depth-first search - DFS (3/4)

- ▶ **Principle:** among open nodes, choose the deepest node (farthest to the root) to expand



Depth-first search - DFS (4/4)

- ▶ **Principle:** among open nodes, choose the deepest node (farthest to the root) to expand



DFS algorithm (1 / 2)

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)

while($O \neq \emptyset$) **do**

1. take the **first node** n from O
2. **if** $n \in G$, **return** (path to n)
3. add $P(n)$ **to the head** of O

return no solution

DFS algorithm (2/2)

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)

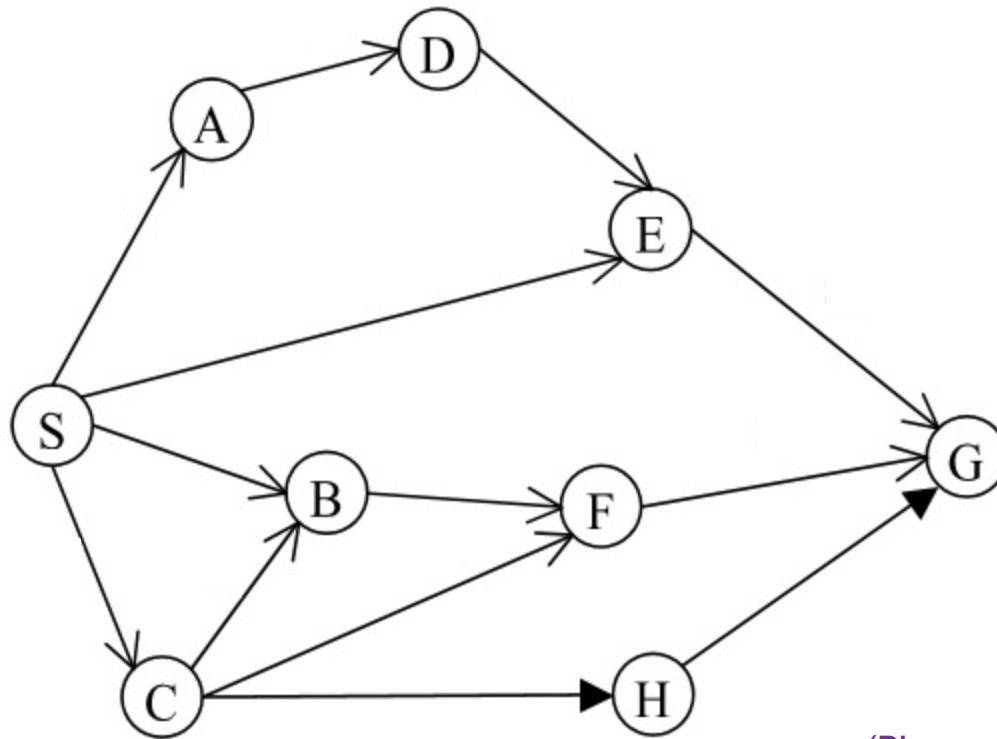
while($O \neq \emptyset$) **do**

1. take the **first node** n from O
2. **if** $n \in G$, **return** (path to n)
3. add $P(n)$ **to the head** of O

return no solution

LIFO data
structure
(Stack)

DFS example (1 / 2)



(Phuong TM, 2016)

DFS example (2/2)

#	Expanded node	Open node list O (Stack)
0		S
1	S	A_S, B_S, C_S, E_S
2	A_S	D_A, B_S, C_S, E_S
3	D_A	E_D, B_S, C_S, E_S
4	E_D	G_E, B_S, C_S, E_S
5	G_E	Goal

Path: $G \leftarrow E \leftarrow D \leftarrow A \leftarrow S$

Depth: 4

Properties of DFS

▶ Completeness?

- No: when the depth of state space is infinite

▶ Optimality?

- No

▶ Time?

- $O(b^m)$: very large if m is greater than d
- If there are many solutions, DFS can be much faster than BFS

▶ Space?

- $O(bm)$: much better than BFS

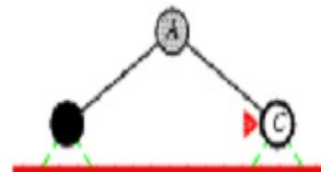
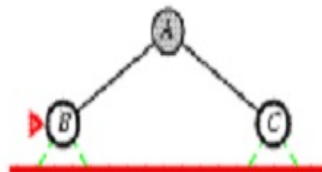
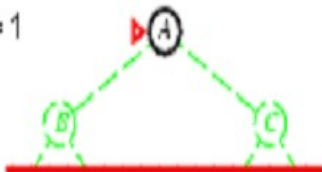
Iterative deepening search–IDS(1 / 3)

- ▶ **Principle:** use DFS but never extend nodes with depth beyond a certain limit. The depth limit will be gradually increased until a solution is found.

Giới hạn = 0

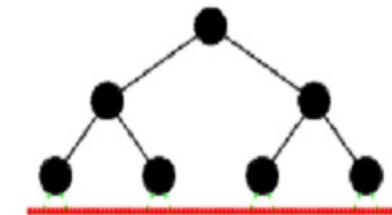
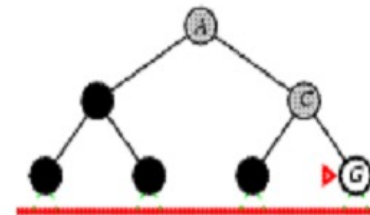
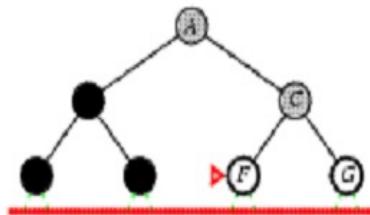
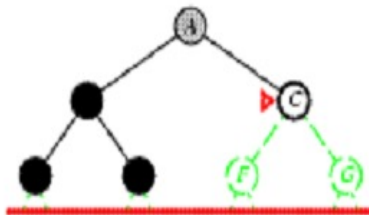
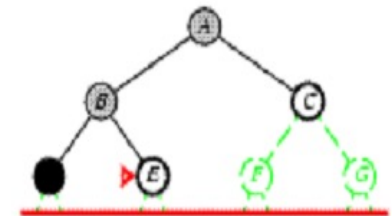
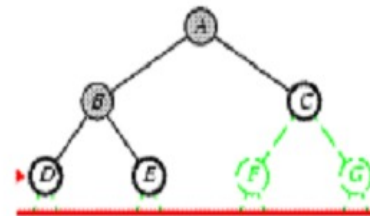


Giới hạn = 1



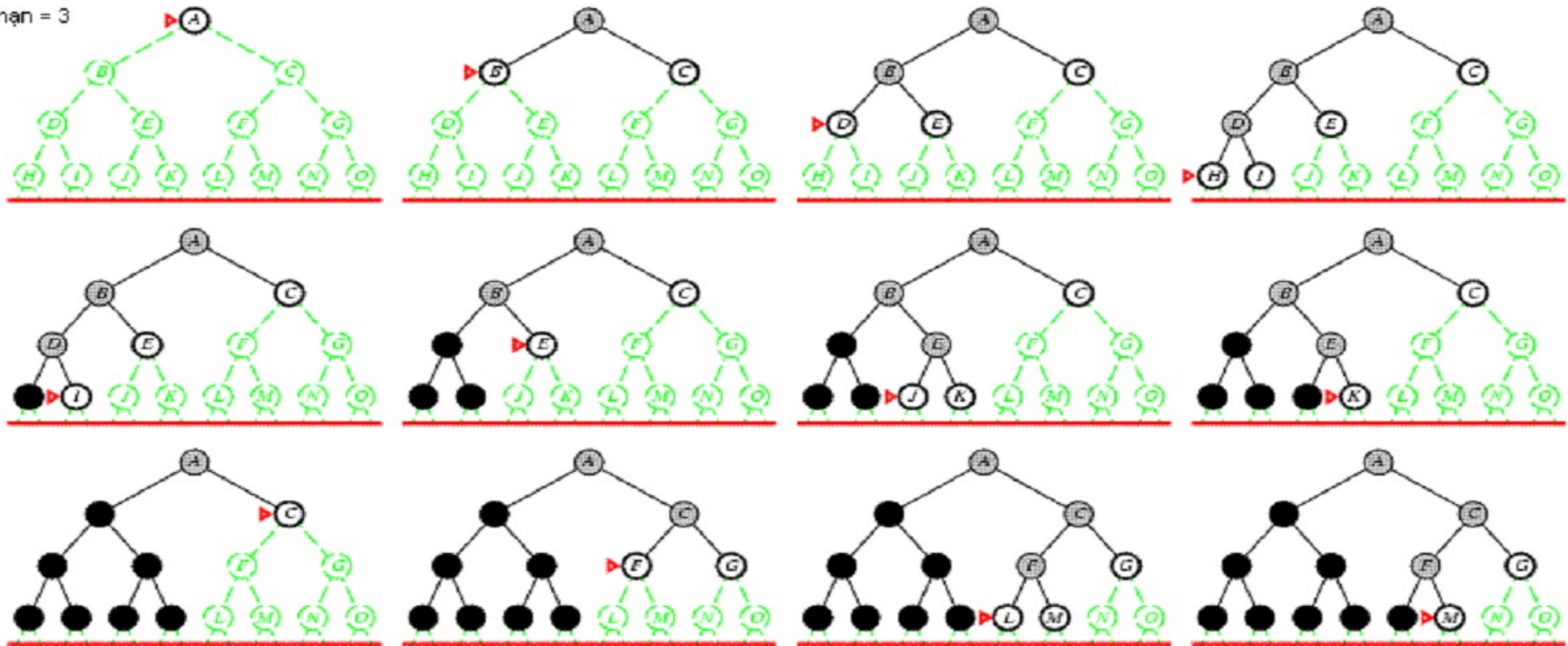
Iterative deepening search–IDS(2/3)

Giới hạn = 2



Iterative deepening search–IDS(3/3)

Giới hạn = 3



IDS algorithm

Search(Q, S, G, P)

(Q : state space, S : initial state, G : goals, P : successor function)

Input: search problem

Output: goal state (path to the goal state)

Initialize: $O \leftarrow S$ (O : the open node list)
 $c = 0$ (current depth)

while (1) **do**

1. **while** ($O \neq \emptyset$) **do**

a. take the first node n from O

b. **if** $n \in G$, **return** (path to n)

c. **if** $depth(n) < c$ **then**

add $P(n)$ to the head of O

2. $c++$; $O = S$

Properties of IDS

▶ Completeness?

- Yes

▶ Optimality?

- Yes: iff there are multiple solutions, IDS can find the solution closest to the root

▶ Space?

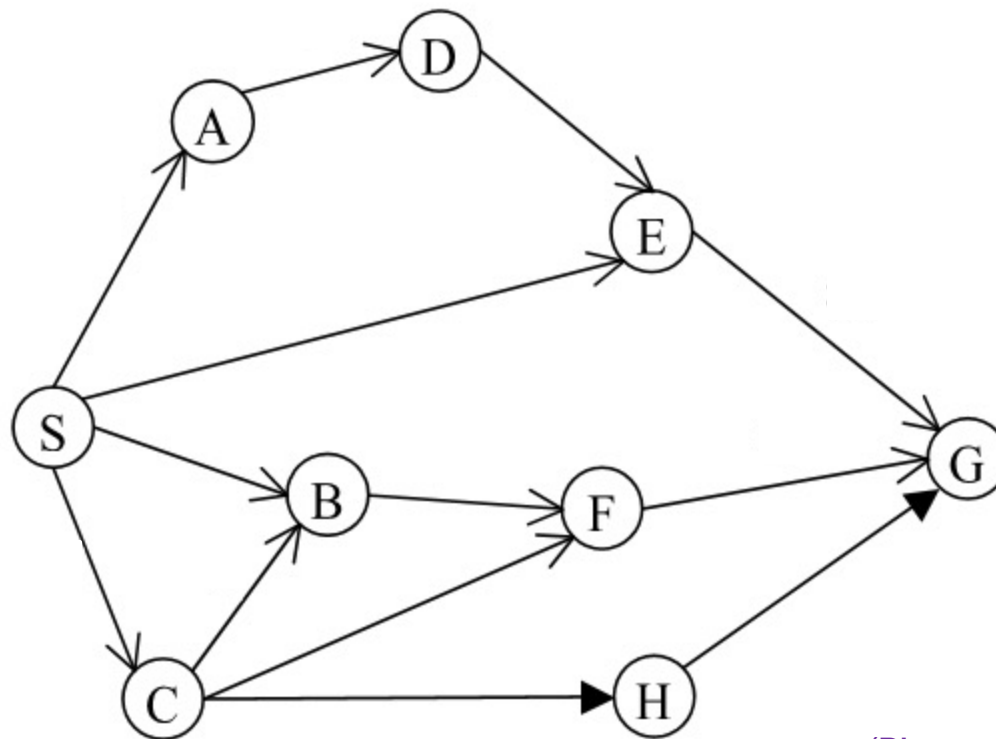
- $O(bd)$: small

▶ Time?

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

Has advantages of both BFS and DFS

IDS example



(Phuong TM, 2016)

Summary

	BFS	UCS	DFS	IDS
Complete?	Yes	Yes	No	Yes
Optimal?	Yes	Yes	No	Yes
Time	$O(b^d)$	$O(b^{\lceil c^*/\epsilon \rceil})$	$O(b^m)$	$O(b^d)$
Space	$O(b^d)$	$O(b^{\lceil c^*/\epsilon \rceil})$	$O(bm)$	$O(bd)$

- ▶ Choose BFS if the branching factor is small
- ▶ Choose DFS if the maximum depth is known in advance and there are multiple goal states
- ▶ Choose IDF if the search tree has a large depth (m)

When to add repeated nodes to the open node list?

▶ BFS

- **No**: adding repeated nodes does not change the order of expanding nodes in the queue; does not change the solution of the problem; may lead to loops.

▶ UCS

- In cases the repeated node has better cost, it will be **added to the list** (if it is already expanded) or **updated to replace the old node** (if it is on the list).

▶ DFS

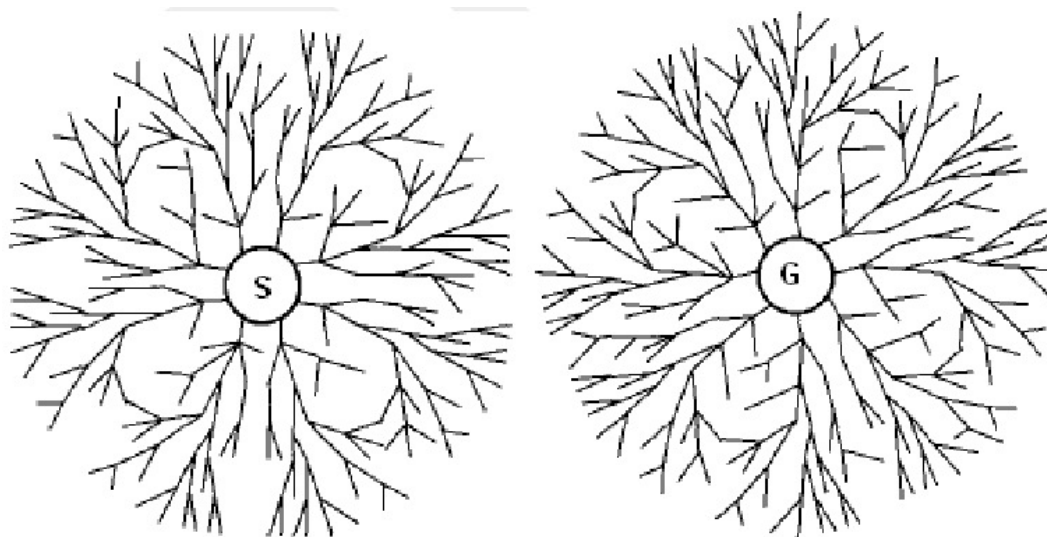
- **Yes**: adding a repeated node to the stack changes the order of expanding nodes in the stack; changes the solution.
- **Do not add expanded nodes** to the stack

▶ IDS:

- **Yes**: to ensure the optimality

Bidirectional search (1/2)

- ▶ **Principle:** simultaneous search from the initial and goal nodes
 - There exist two search trees, one rooted at the initial node and the other rooted at the goal node
 - Search ends when a leaf of one tree matches a leaf of the other
- ▶ Illustration of search trees



(Phuong TM, 2016)

Bidirectional search (2/2)

► Note

- Need to use **BFS**
 - DFS may not yield a solution if two search trees grow along two branches that do not match each other
- Functions
 - $P(x)$: set of child nodes of x
 - $D(x)$: set of father nodes of x

► Properties

- Node matching is time consuming (b^d nodes for each tree)
- Computational complexity $O(b^{d/2})$
 - Number of expanded nodes is significantly reduced