

# UIT2504 Artificial Intelligence

## Basic Search Strategies

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# State-Space Approach

- Given an **initial state**, and a **set of actions**, “**search tree**” can be visualized
- The number of possible actions determine the **branching factor** of the tree
- **State space** is the set of all states that are reachable from the initial state using a sequence of actions
- Which states will be leaf nodes in the search tree? — all **goal states** will be leaf nodes — there can also be “**dead-end**” states where no action is possible
- Where is the **solution** in the search tree? — normally a path from initial state to a goal state
- In some cases, goal state itself is a solution!

# Problem Formulation

- A problem may be formulated in the state-space approach by defining the following four components:
  - Set of states, with the **initial state**
  - Set of possible actions, which may be abstracted by a **successor function**  $S$ . For any state  $x$ ,  $S(x)$  is the set of states reachable from  $x$  using an action
  - **Goal test** — a function that returns true if a given state is a goal state
  - **Path cost** — sum of all the costs of actions from initial state to a goal state

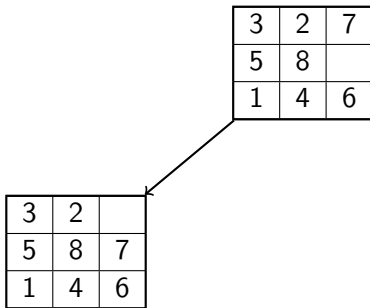
# Illustration: 8-Puzzle

- *State*: Arrangement of eight tiles on a  $3 \times 3$  grid — *Initial state* is some random arrangement of eight tiles on nine squares
- *Actions*: blank moves left, right, up, or down — branching factor of 4
- *Goal test*: state matches the given target configuration
- *Path cost*: unit cost for each action

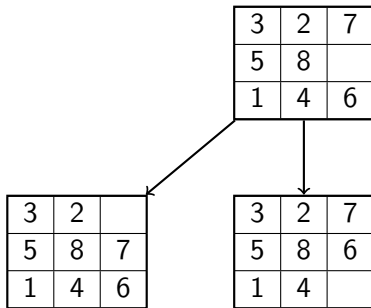
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|   |   |   |
|---|---|---|
| 3 | 2 | 7 |
| 5 | 8 |   |
| 1 | 4 | 6 |

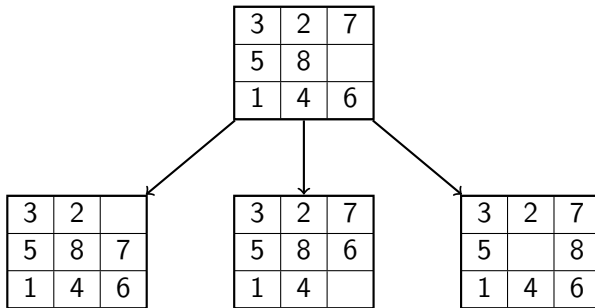
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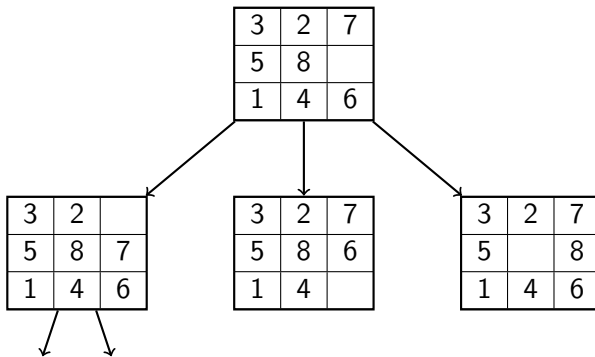


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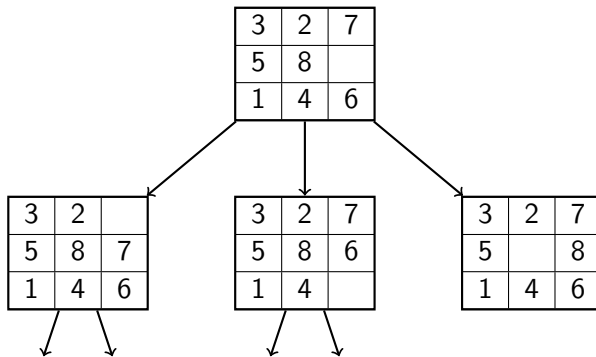




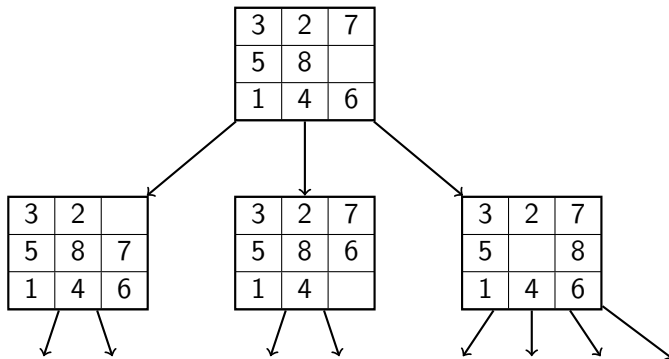
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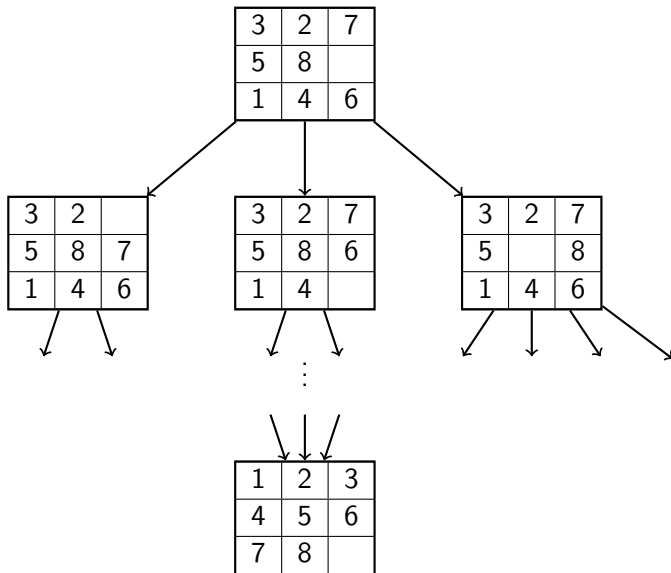
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  - Return failure if the working set is empty
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  - Else, expand  $x$  and add the successor states  $S(x)$  to the working set

- Uninformed:
  - Breadth-First
  - Depth-First
  - Depth-Limited
  - Bi-Directional Search
- Informed (Heuristics):
  - Best-first Greedy
  - $A^*$
  - Local Search Strategies
- Constraint Satisfaction

# Performance Measures

- Completeness
- Time Complexity
- Space Complexity
- Optimality

# Questions?

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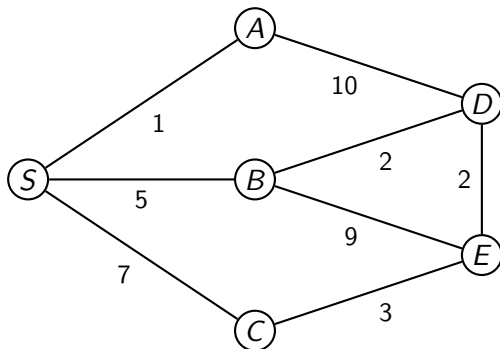
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- All nodes at depth  $d$  are expanded before any node at depth  $d + 1$

# BFS: Illustration

- How does it work on our toy example?



Find a route from  $S$  to  $E$



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- From tree point of view, nodes are examined in the level order
- Each and every path is progressively examined!

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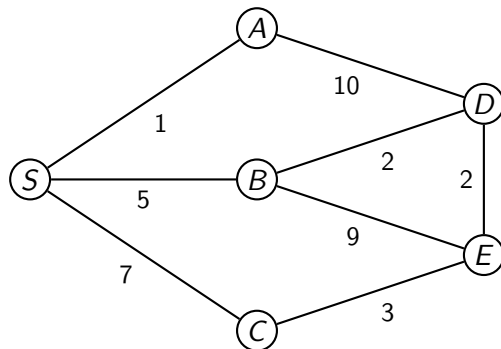
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- Working set (fringe) is a partial order arranged in ascending order of cost (minheap)
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- Does this remind you of any graph algorithm that you have learnt?

- How does it work on our toy example?



Find a route from  $S$  to  $E$

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- Selectively switches between the paths!

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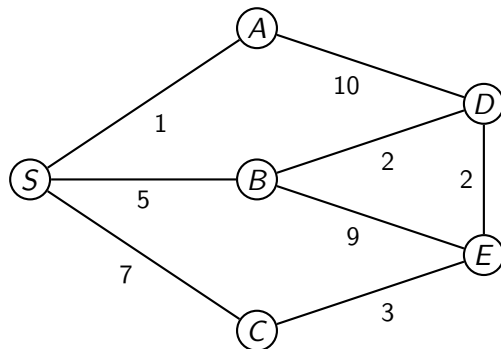
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- Preference is given to newly generated nodes

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- Can you suggest some modification to DFS to make it complete?