

# AI1 & AIML - DFS, Variations and Informed Search

Dr Leonardo Stella

## Aims of the Session

This session aims to help you:

Explain the steps involved in Depth-First Search and its variations.
 Be able to apply these algorithms to solve search problems

Understand the concept of a heuristic function

 Analyse the performance of A\* and apply the algorithm to solve search problems

#### Overview

- Recap: Search Problem Formulation and BFS
- Depth-First Search and Variations
- Informed Search
- A\* Algorithm

- Activity. Explain what each of the following terms mean:
  - Observable
  - Discrete
  - Known
  - Deterministic

- Activity. Explain what each of the following terms mean:
  - **Observable**, i.e., the agent is able to know the current state
  - **Discrete**, i.e., there are only finitely many actions at any state
  - Known, i.e., the agent can determine which states are reached by which action
  - Deterministic, i.e., each action has exactly one outcome

Activity. Explain what the main components of a search problem are:

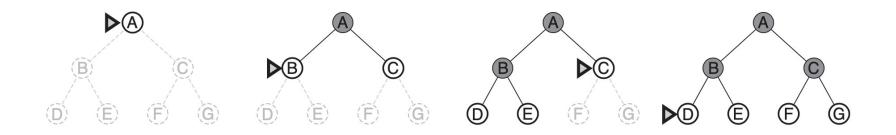
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  - Initial state, Actions, Transition model, Goal test and Path cost

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The first three components together define the state space of the problem, in the form of a directed graph or network. A path in the state space is a sequence of states connected by a sequence of actions

#### Breadth-First Search

- Breadth-First Search steps (see Fig. 3.12):
  - Expand the shallowest node in the frontier
  - **Do not add** children in the frontier if the node is already in the frontier or in the list of visited nodes (to avoid loopy paths)
  - Stop when a goal node is added to the frontier



#### **BFS** - Performance

Let us evaluate the performance of BFS:

- Completeness: if the goal node is at some finite depth d, then the BFS algorithm is complete as it will find it (given that b is finite)
- Optimality: BFS is optimal if the path cost is a nondecreasing function of the depth of the node (e.g., all actions have the same cost)
- **Time complexity**:  $O(b^d)$ , assuming a uniform tree where each node has b successors, we generate  $b + b^2 + \cdots + b^d = O(b^d)$
- **Space complexity**:  $O(b^d)$ , if we store all expanded nodes, we have  $O(b^{d-1})$  explored nodes in memory and  $O(b^d)$  in the frontier

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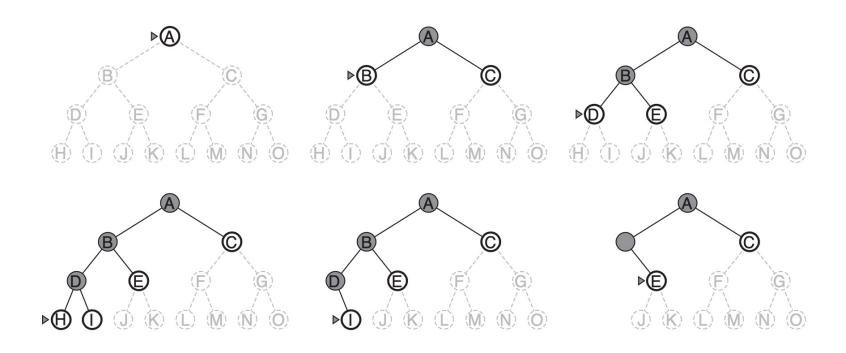
- Depth-First search is another common search strategy:
  - The root node is expanded first
  - Then, the first (or one at random) successor of the root node is expanded
  - Then, the deepest node in the current frontier is expanded

 This is equivalent to expanding the deepest unexpanded node in the frontier; simply use a stack (LIFO) for expansion

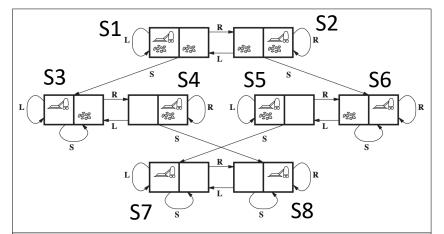
Basically, the most recently generated node is chosen for expansion

- Depth-First Search steps:
  - Expand the deepest node in the frontier
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# Depth-First Search (see Fig. 3.16)



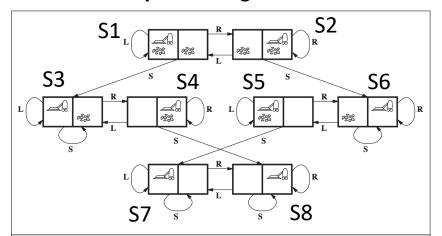
- Activity. Let us apply DFS to the Vacuum World example from the book:
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**S3** 

Stop when a goal node is visited

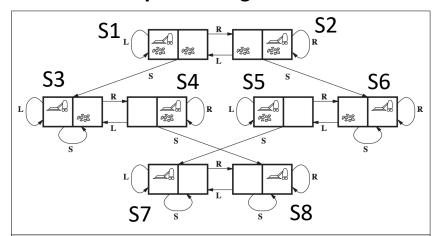


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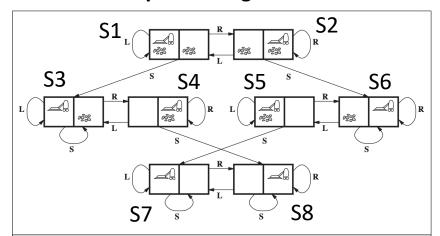
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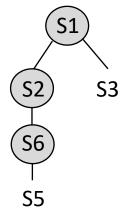
**S6** 

Stop when a goal node is visited



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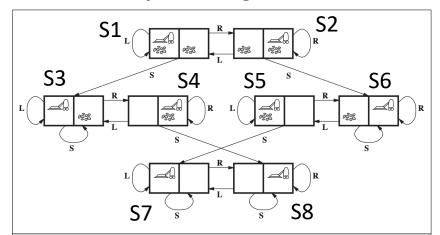


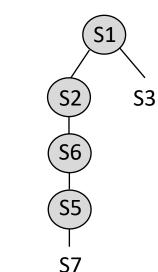


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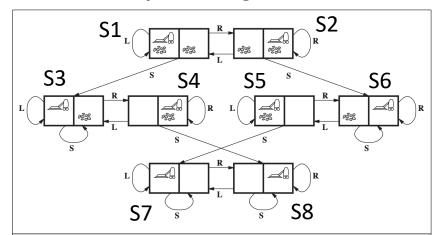
list of visited nodes (to avoid loopy paths)

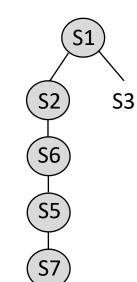
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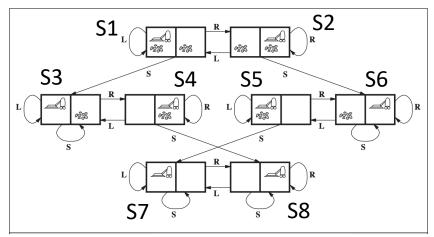
Solution:

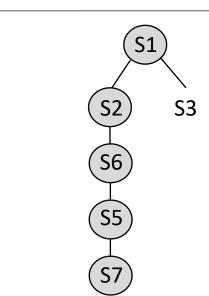
R, S, L, S

Cost of the solution:

$$1 + 1 + 1 + 1 = 4$$

Order of nodes visitedS1, S2, S6, S5, S7





#### DFS - Performance

Activity. Let us evaluate the performance of DFS:

- Completeness:
- Optimality:
- Time complexity:
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- Time complexity:
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- Optimality: DFS is not optimal as it can expand a left subtree when the goal node is in the first level of the right subtree
- **Time complexity**:  $O(b^m)$ , as it depends on the maximum length of the path in the search space (in general m can be much larger than d)
- **Space complexity**:  $O(b^m)$ , as we store all the nodes from each path from the root node to the leaf node

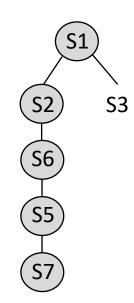
## Depth-First Search - Variations

- Depth-First Search comes with several issues
  - Not optimal
  - High time complexity
  - High space complexity
- DFS with less memory usage (saving space complexity)

Depth-Limited Search

Imagine we have a tree similar the one in the example

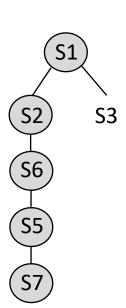
Now, S7 is not a goal node and it has no children



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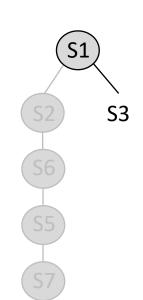
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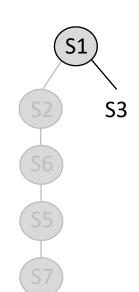
- The next step of the algorithm would be to expand S3
- Since we explored all the left subtree, we can remove it from memory



■ This would reduce the space complexity to O(bm)

 We need to store a single path along with the siblings for each node on the path

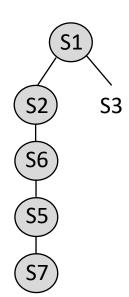
lacktriangle Recall that b is the branching factor and m is the maximum depth of the search tree



# Depth-Limited Search

■ The issue related to depth-first search in infinite state spaces can be mitigated by providing a depth limit  $\ell$ 

This approach is called depth-limited search

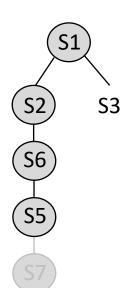


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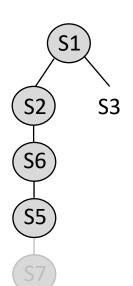
• For  $\ell = 3$ , we would have



# Depth-Limited Search

This adds an additional source of incompleteness if we choose  $\ell < d$ , namely, the shallowest goal is beyond the depth limit

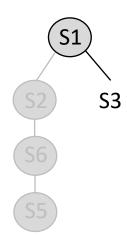
- This approach is nonoptimal also in the case  $\ell > d$
- Time complexity is  $O(b^{\ell})$



# Depth-Limited Search – Less Memory Usage

- As before, we can remove the explored paths from memory after we have reached the depth limit  $\ell$ 

• Space complexity is  $O(b\ell)$ 



# Comparing Uninformed Search Strategies

Criterion / Algorithm	Breadth-First	Depth-First	Depth-First (less memory)	Depth-Limited (less memory)
Completeness	Yes*	Yes***	Yes***	Yes, if $\ell \geq d$
Optimality	Yes**	No	No	No
Time	$O(b^d)$	$O(b^m)$	$O(b^m)$	$O(b^\ell)$
Space	$O(b^d)$	$O(b^m)$	O(bm)	$O(b\ell)$

<sup>\*</sup> If b is finite

<sup>\*\*</sup> If the path cost is a nondecreasing function of the depth of the node (e.g., all actions have the same cost)

<sup>\*\*\*</sup> If the search space is finite (also, loopy paths are removed)

## Summary

- Depth-First Search is a search algorithm that expands the nodes in the frontier starting from the deepest, similar to a stack (LIFO)
- This algorithm is complete (for finite search space), but not optimal; also, it has high time complexity and space complexity  $O(b^m)$
- Depth-First Search can be improved in terms of its time and space complexity
- Depth-First Search with less memory usage only keeps in memory the current path and the siblings of the nodes. Depth-Limited Search is another variation, where a depth limit is specified; this adds an additional source of incompleteness

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## Informed Search Strategies

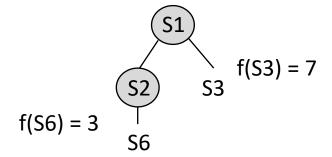
 Informed search strategies use problem-specific knowledge beyond the definition of the problem itself

 Informed search strategies can find solutions more efficiently compared to uninformed search

# **Informed Search Strategies**

■ The general approach, called **best-first search**, is to determine which node to expand based on an **evaluation function** f(n):  $node \rightarrow cost\ estimate$ 

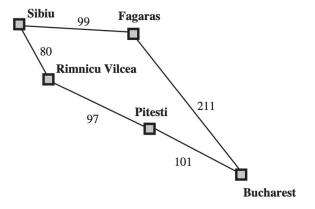
 This function acts as a cost estimate: the node with the lowest cost is the one that is expanded next



- The evaluation function f(n) for most best-first search algorithms includes a **heuristic function** as a component: h(n) = estimated cost of the cheapest path from node <math>n to a goal node
- Heuristic functions are the most common form in which new knowledge is given to the search algorithm. If n is a goal node, then h(n) = 0

 A heuristic can be a rule of thumb, common knowledge; it is quick to compute, but not guaranteed to work (nor to yield optimal solutions)

Consider the problem to find the shortest path to Bucharest in Romania



- We can use the straight-line distance heuristic, denoted by  $h_{SLD}$
- This is a useful heuristic as it is correlated with actual road distances

Consider the problem to find the shortest path to Bucharest in Romania



- The straight-line distances  $h_{SLD}$  are shown in the table above
- For example, the SLD from Sibiu would be 253

Consider the problem to find the shortest path to Bucharest in Romania



- If we use  $f(n) = h_{SLD}(n)$ , then from Sibiu we expand Fagaras
- This is because Fagaras has SLD 176, while Rimnicu Vilcea 193

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• When f(n) = h(n), we call this strategy **Greedy Best-First Search** 

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■ The most widely known informed search strategy is A\*

This search strategy evaluates nodes using the following cost function

$$f(n) = g(n) + h(n)$$

where g(n) is the cost to reach the node and h(n) is the heuristic from the node to the goal

 $\blacksquare$  This is equivalent to the cost of the cheapest solution through node n

Consider the problem to find the shortest path to Bucharest in Romania



• Activity. Consider the above problem where Sibiu is the initial state. Calculate f(n) to choose which node to expand, starting with Fagaras f(Fagaras) = g(Fagaras) + h(Fagaras)

Consider the problem to find the shortest path to Bucharest in Romania



Activity. Consider the above problem where Sibiu is the initial state. Calculate f(n) to choose which node to expand, starting with Fagaras f(Fagaras) = 99 + 176 = 275

Consider the problem to find the shortest path to Bucharest in Romania



• Activity. Consider the above problem where Sibiu is the initial state. Calculate f(n) to choose which node to expand:

$$f(Rimnicu\ Vilcea) =$$

Consider the problem to find the shortest path to Bucharest in Romania

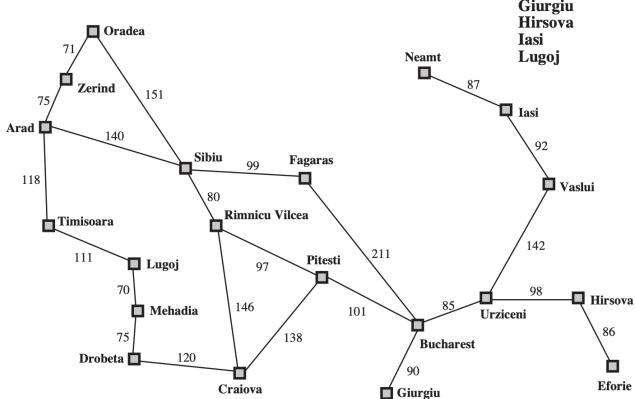


• Activity. Consider the above problem where Sibiu is the initial state. Calculate f(n) to choose which node to expand:

$$f(Rimnicu\ Vilcea) = 80 + 193 = 273$$

## A\* Search - Algorithm

- A\* search algorithm:
  - **Expand** the node in the frontier with smallest cost f(n) = g(n) + h(n)
  - Do not add children in the frontier if the node is in the list of visited nodes (to avoid loopy paths)
  - If the state of a given child is in the frontier
    - If the frontier node has a larger g(n), place the child into the frontier and remove the node with larger g(n) from the frontier
  - Stop when a goal node is visited



Mehadia **Arad** 366 241 **Bucharest** Neamt 234 0 Craiova 160 **Oradea** 380 **Drobeta** 242 **Pitesti** 100 **Eforie** Rimnicu Vilcea 161 193 **Fagaras** Sibiu 176 253 **Timisoara** 329 77

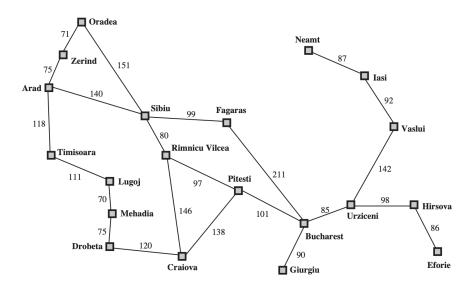
 Fagaras
 176
 Sibiu
 253

 Giurgiu
 77
 Timisoara
 329

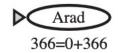
 Hirsova
 151
 Urziceni
 80

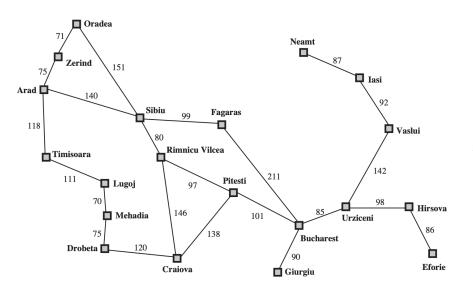
 Iasi
 226
 Vaslui
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 Lugoj
 244
 Zerind
 374

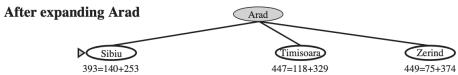


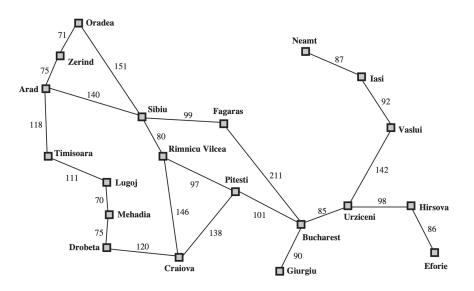
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<b>Bucharest</b>	0	Neamt	234
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Giurgiu	77	Timisoara	329
Hirsova	151	Urziceni	80
Iasi	226	Vaslui	199
Lugoj	244	Zerind	374



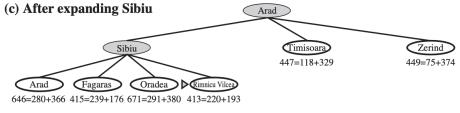


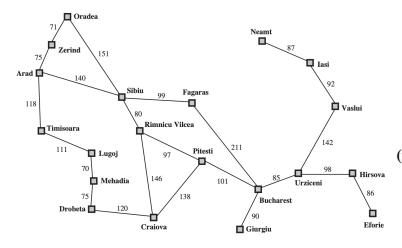
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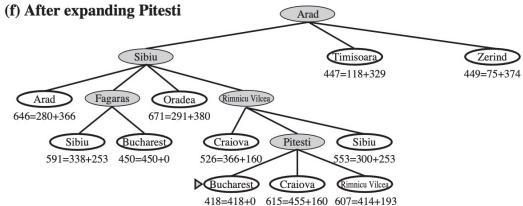


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# A\* Search - Completeness and Optimality

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■ The A\* search is **complete** and **optimal** if h(n) is consistent

 Definition: a heuristic is said to be consistent (or monotone), if the estimate is always no greater than the estimated distance from any neighbouring node to the goal, plus the cost of reaching that neighbour.

$$h(n) \le cost(n, n') + h(n')$$

# A\* Search - Time and Space Complexity

- The number of states for the A\* search is **exponential** in the length of the solution, namely, for constant step costs:  $O(b^{\epsilon d})$
- When  $h^*$  is the actual cost from root node to goal node,  $\epsilon = \frac{(h^* h)}{h^*}$  is the relative error

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 Space is the main issue with A\*, as it keeps all generated nodes in memory, therefore A\* is not suitable for many large-scale problems

## A\* Search - Summary

Let us summarise the performance of the A\* search algorithm

- Completeness: if the heuristic h(n) is consistent, then the A\* algorithm is complete
- Optimality: if the heuristic h(n) is consistent, A\* is optimal

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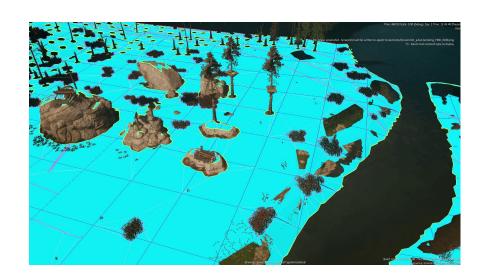
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- **Time complexity**:  $O(b^{\epsilon d})$ , where  $\epsilon$  is the relative error of the heuristic
- **Space complexity**:  $O(b^d)$ , since we keep in memory all expanded nodes and all nodes in the frontier

# A\* - Applications

- A\* has a large number of applications
- In practice, the most common ones are in games and in robotics



R.O.B.O.T. Comics



"HIS PATH-PLANNING MAY BE SUB-OPTIMAL, BUT IT'S GOT FLAIR."

## Summary

A\* is complete and optimal, given a consistent heuristic

 However, A\* has typically high time/space complexity, regardless of the heuristic chosen

 Heuristics have a considerable impact on the performance of informed search algorithms, and they can drastically reduce the time and space complexity in comparison to uninformed search algorithms

#### Aims of the Session

You should now be able to:

Explain the steps involved in Depth-First Search and its variations.
 Be able to apply these algorithms to solve search problems

Understand the concept of a heuristic function

 Analyse the performance of A\* and apply the algorithm to solve search problems

#### References

- Russell, A. S., and Norvig, P. (2010), Artificial Intelligence A Modern Approach, 3<sup>rd</sup> Edition. Prentice Hall.
  - Chapter 3 Solving Problems by Searching (Section 3.4.3 up to 3.4.4, Section 3.4.7, Section 3.5 up to 3.5.2)