# Solutions to Exercises - Lecture 3 Intelligent Systems Programming

# **Exercise 1**

#### a)

Portion of the state space is shown in Figure 1.

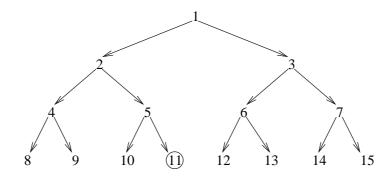


Figure 1: Portion of the state space for states 1 to 15 in Exercise 1.

b)

- DLS:  $\{1, 2, 4, 8, 9, 5, 10, 11\}$
- IDS: {1; 1, 2, 3; 1, 2, 4, 5, 3, 6, 7; 1, 2, 4, 8, 9, 5, 10, 11}

c)

Branching factor is one. There is only one action that can be taken when going backwards.

d)

Given a starting problem with result function result, starting state  $s_0=1$  and some goal state  $s_G$ , define a new problem, with "inverse" result function result', such that:  $result'(n)=\lfloor n/2\rfloor$ . The initial state is an old goal state  $s_0'=s_G$  and a goal state is  $s_G'=s_0=1$ . To implement required algorithm it suffices to run a tree-search with BFS over the new problem.

Since branching factor is one, in every step we expand only one state s with only one result s' = result'(s) which is at smaller depth than s. Hence, the number of generated nodes is equal to the depth of the goal state. Since the tree is a balanced binary tree, we have that the depth is O(log(k)). Thus, the complexity of our algorithm is O(log(k)).

# **Exercise 2**

a)

We represent each of n disks with a number from  $S = \{1, \ldots, n\}$  assuming that smaller disks are identified with smaller numbers, i.e., the i-th disk is smaller than the j-th disk if i < j. A state is defined with a partition of S into three sets  $(S_1, S_2, S_3)$  (i.e.,  $\bigcup_{i=1}^3 S_i = S$ , and  $S_i \cap S_j = \emptyset$  for all  $i \neq j$ ) where a set  $S_k$  represents disks that are on the k-th peg. Once a set  $S_k$  is given, we know exactly how the disks are put on the k-th peg, since there is only one way to order them (smaller on top). In particular, the topmost disk in set  $S_k$  is  $min\{a \mid a \in S_k\}$ , which we denote simply as  $minS_k$ .

We have six actions M(1,2), M(1,3), M(2,1), M(2,3), M(3,1), M(3,2), where M(i,j) denotes an action of moving a smallest disk from the i-th peg, and putting it on the j-th peg. For each state only some of the actions might be legal, i.e., the topmost disk on the i-th peg must be smaller than topmost disk on the j-th peg  $(minS_i < minS_j)$ . The *Initial state* is given by  $S_1 = \{1, \ldots, n\}, S_2 = \emptyset, S_3 = \emptyset$ .

The actions function for each state  $(S_1, S_2, S_3)$  returns all legal action-state pairs  $(M(i,j), (S_1', S_2', S_3'))$ , where the action M(i,j) is legal. For each such action M(i,j), in the results function gives the resulting state in which the two affected pegs are  $S_i' = S_i \setminus \{minS_i\}$ ,  $S_j' = S_j \cup \{minS_i\}$  while the remaining one stays the same.

Goal test is given by checking whether  $S_3 = \{1, 2, ..., n\}$ . Path cost is a positive constant, for example c(s, a, s') = 1.

#### b)

The state space is shown in Figure 2.

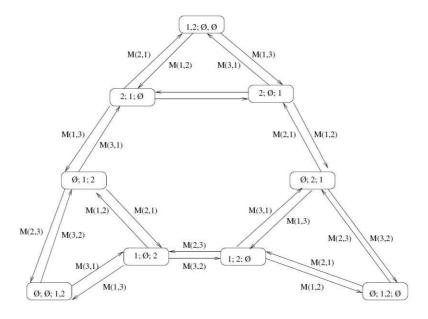


Figure 2: State space of Towers of Hanoi with n=2.  $\emptyset$  denotes a peg without disks.

## c)

Notice the difference between legal states (i.e., all states satisfying our definition of a state) and reachable states (i.e., states we can reach by executing some sequence of actions from the starting state). Reachable states are in general a subset of legal states. Since we are assuming all legal states are reachable, we need only to count the number of states  $(S_1, S_2, S_3)$  that satisfy our definition from a). The number of all legal configurations is  $3^n$  since for every disk we can choose any of the three pegs to put it on.

## Exercise 3

# a)

The nodes in the fringe of the A\*-algorithm when solving the specified problem are given in the table below. Notice that the fringe is a priority queue, and that the

nodes in it are therefore ordered in accordance to their f value.

Iteration	Fringe
0	$\langle 244, 0, 244, L \rangle$
1	$\langle 311, 70, 241, M \rangle, \langle 440, 111, 329, T \rangle$
2	$\langle 387, 145, 242, D \rangle, \langle 440, 111, 329, T \rangle$
3	$\langle 425, 265, 160, C \rangle, \langle 440, 111, 329, T \rangle$
4	$\langle 440, 111, 329, T \rangle, \langle 503, 403, 100, P \rangle, \langle 604, 411, 193, RV \rangle$
5	$\langle 503, 403, 100, P \rangle, \langle 595, 229, 366, A \rangle, \langle 604, 411, 193, RV \rangle$
6	$\langle 595, 229, 366, A \rangle, \langle 604, 411, 193, RV \rangle, \langle 504, 504, 0, B \rangle$

#### b)

A\* returns Solution(goal) where goal is a goal node. The mentioned algorithm Solution traces the path from goal back to the initial node and returns the found path.

In this case,  $\mathbf{A}^*$  therefore returns the path:  $L \to M \to D \to C \to P \to B$ .

# **Exercise 4**

#### **a**)

For w=1 we have greedy best first search, for w=0.5 we have  $A^*$  search and for w=0 we have uniform cost search.

#### b)

Notice that we can multiply f(n) with a positive number without changing the behavior of the algorithm. Now multiply f(n) with  $\frac{1}{(1-w)}$  which gives  $\frac{1}{(1-w)}f(n)=g(n)+\frac{w}{1-w}h(n)$ . The heuristic function  $\frac{w}{1-w}h(n)$  will be admissible for  $\frac{w}{1-w}\leq 1$  since h(n) is admissible. This is the case for  $w\leq 0.5$ .

# **Exercise 5**

## a)

Let k denote the number of edges in  $p^*(n)$ ,  $k = |p^*(n)|$ .

- If k = 0, a state n is a goal state  $s_G$ . Therefore h(n) = 0 according to the definition of heuristic function h(n). But also,  $h^*(n) = 0$  since obviously the cheapest path to goal has length 0. Therefore it holds  $h(n) \le h^*(n)$ .
- Assume that the claim holds for paths with k edges. Let us show that the statement also holds for paths with k+1 edges. Let  $p^*(n) = \{n, n_1, \ldots, n_{k+1}\}$  be the optimal path with k+1 edges where  $n_{k+1}$  is the goal state. A cost of each edge  $c(n_i, n_{i+1})$  is the standard cost of executing an action from a state  $n_i$  that leads to a state  $n_{i+1}$ .

Since the statement holds for paths with k edges, it also holds for state  $n_1$ . Namely, the optimal path from  $n_1$  to goal  $n_{k+1}$ ,  $p_1^*(n_1) = \{n_1, \ldots, n_{k+1}\}$ , has k edges. Therefore:

$$h(n_1) \le h^*(n_1) \tag{1}$$

Also, since  $p^*(n_1)$  is the optimal path from  $n_1$ , its length  $\sum_{i=1}^k c(n_i, n_{i+1})$  is equal to  $h^*(n_1)$ . Since the similar holds for  $h^*(n)$  it follows:

$$h^*(n) = c(n, n_1) + h^*(n_1)$$
(2)

Now we have everything we need to prove  $h(n) \leq h^*(n)$ . Since, h is consistent heuristic it holds  $h(n) \leq c(n, n_1) + h(n_1) \leq^{(1)} c(n, n_1) + h^*(n_1) =^{(2)} h^*(n)$ . This proves the induction step.

• Since the statement holds for k = 0, and from the fact that if the statement holds for k we can show that it also holds for k+1, according to the principle of mathematical induction, the statement holds for all  $k \in \mathbb{N}$ .

#### b)

Assume that h is not admissible. Then for some state  $n_0$  it holds  $h(n_0) > h^*(n_0)$ . However, relation > makes sense only if both  $h(n_0)$  and  $h^*(n_0)$  are finite. But this means that a goal is reachable from  $n_0$  in finite number of steps, i.e., there is an optimal path  $p^*(n_0)$ . Then according to a) it must hold  $h(n_0) \le h^*(n_0)$  which contradicts the initial assumption. Therefore h is admissible.