# CSCE 790-002: Deep Reinforcement Learning and Search Homework 4

Due: 12/1/2020

## Installation

We will be using the same conda environment as in Homework 3.

The entire GitHub repository should be downloaded again as changes were made to other files. You can download it with the green "Code" button and click "Download ZIP".

## A\* Search

For this homework, you will implement  $A^*$  search and, with only a one-line modification, you will also implement weighted  $A^*$  search, greedy best-first search, and uniform cost search. The  $A^*$  search algorithm is outlined in Algorithm 1.

For a visualization of a working implementation, see Blackboard under Course Content/Homework/Homework 4.

At each iteration, a node is popped from OPEN that has the highest priority. The priority is determined by the cost f where the node with the lowest cost has the highest priority. The cost f is computed as  $f(n) = \lambda_g g(n) + \lambda_h h(n)$ . g(n) is the cost of getting to the start node to node n, h(n) is the estimated cost of getting from node n to the goal node and  $\lambda_g$  and  $\lambda_h$  are weights.

```
A* search is obtained by setting \lambda_g=1 and \lambda_h=1.
Weighted A* search is obtained by varying either \lambda_g or \lambda_h.
Greedy best-first search is obtained by setting \lambda_g=0 and \lambda_h=1.
Uniform cost search is obtained by setting \lambda_g=1 and \lambda_h=0.
```

To run each of the methods use:

#### A\* Search

```
python run_astar.py --map maps/map1.txt --g_weight 1.0 --h_weight 1.0
```

### Weighted A\* Search

```
python run_astar.py --map maps/map1.txt --g_weight 0.1 --h_weight 1.0 python run_astar.py --map maps/map1.txt --g_weight 0.5 --h_weight 1.0
```

#### Greedy Best-First Search

```
python run_astar.py --map maps/map1.txt --g_weight 0.0 --h_weight 1.0
```

#### Uniform Cost Search

```
python run_astar.py --map maps/map1.txt --g_weight 1.0 --h_weight 0.0
```

where 
$$--g_{\text{weight}}$$
 is  $\lambda_q$  and  $--h_{\text{weight}}$   $\lambda_h$ 

## Algorithm 1 A\* Search

```
1: procedure A* SEARCH(s_0, h_\theta, \lambda_q, \lambda_h)
       Initialize OPEN, CLOSED
 2:
                                                                      > priority queue and set of expanded nodes
                                                  ⊳ have not yet found the node associated with the goal state
       n_q = None
 3:
 4:
        Create node n_0 with state s_0
        push n_0 to OPEN
5:
       while n_q is None do
 6:
           if OPEN is Empty then
 7:
               return failure
8:
           end if
9:
           pop highest priority node n from OPEN and put it in CLOSED
10:
           if n is a goal node then
11:
12:
               n_q = n
           end if
13:
14:
           n_{children} = \text{expand\_node}(n.state, a)
           n_{children} = \text{remove\_nodes\_in\_open\_or\_closed}(n_{children})
15:
           costs = compute\_costs(n_{children}, h_{\theta}, \lambda_{q}, \lambda_{h})
16:
           push nodes n-children to OPEN
17:
        end while
18:
        return get_soln(n_a)
19:
20: end procedure
```

The Node class has the following properties:

state: the state associated with the node

path\_cost: the cost to go from the start state to the current state

parent\_action: the action taken from the parent node to get to the current node

parent: the parent node

## Expand Nodes (35 pts)

Implement the expand\_node method. This method will take a parent node and return a list of its child nodes. These child nodes are the result of taking every possible action from the state associated with the parent node (parent\_node.state).

Instantiating a node is done with Node(state, path\_cost, action, parent\_node).

Use self.env.get\_actions() to get all possible actions.

Use self.get\_next\_state\_and\_transition\_cost(node.state, action) to get the resulting state and transition cost of taking the given action in the state associated with the given node. You can get the correct path\_cost of the child node by adding the path\_cost of the parent node to the transition cost.

## Compute Costs (35 pts)

Implement the compute\_cost function.

The heuristic\_fn takes a list of states as an input and returns a np.array of the computed heuristics. The compute\_cost function should return the costs (a list of floats) where the cost of a node n is computed as  $f(n) = \lambda_q g(n) + \lambda_h h(n)$ 

# Get Solution (30 pts)

Implement the get\_soln function.

This function should return the sequence of actions needed to get to the given node from the start node. The return type should be a list of integers.

Use node.parent and  $node.parent_action$ . Keep in mind that, for the start node, node.parent is None and  $node.parent_action$  is None.

## What to Turn In

Turn in your implementation of assignments\_code/assignment4.py.