Task 3. Reinforcement Learning

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1 Introduction

2 A* path finding

The A* algorithm can find paths optimally in a graph based on a heuristic. It works under the assumption that the distance heuristic will never overestimate the cost of the path and the path cost will not decrease as we travel through it. In a maze world, where the movement possibilities are north, south, east and west, with fixed positive costs. Using a Manhattan distance for the actual path cost and a euclidean distance to compute the heuristic will fit the assumptions.

In Algorithm 1 an overview of a general implementation of the A* in pseudocode is given. As stated in the previous paragraph, in our case, graph.cost() and heuristic() returns respectively, the Manhattan and the euclidean distance between two nodes. An example output with a small labyrinth where a path is successfully found is shown in Figure 1.

- 3 Q learning in noughts and crosses
- 4 Q learning in BlackJack
- 5 Results and conclusions

Algorithm 1: A*

```
Data: goal goal position, start start position, graph graph with the tiles in the map.
Result: path path from start to goal, cost total cost of the path.
frontier = PriorityQueue();
frontier.put(start, 0);
came\_from = \{\};
cost\_so\_far = \{\};
came\_from[start] = None;
cost\_so\_far[start] = 0;
while not frontier.empty() do
   current = frontier.get();
   if current == goal then
       break;
   end
   for next in graph.neighbors(current) do
       new_cost = cost_so_far[current] + graph.cost(current, next);
       if next not in cost_so_far or new_cost < cost_so_far/next/ then
           cost\_so\_far[next] = new\_cost;
           priority = new\_cost + heuristic(goal, next);
          frontier.put(next, priority);
           came\_from[next] = current;
       end
   end
end
path = getPath(came\_from);
goal = cost\_so\_far[current];
```

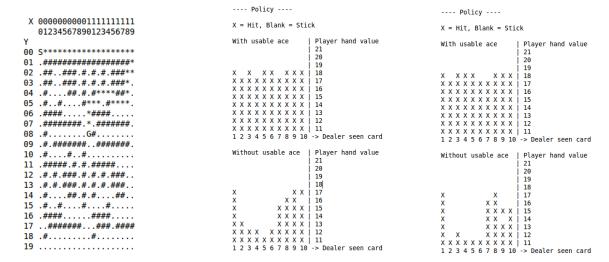


Figure 1: Path finding example, S is the start position, G is the goal position, G are empty spaces and G are walls.

Figure 2: Blackjack policy example, with 20% exploration rate after 100000 games.

Figure 3: Blackjack policy example, with 40% exploration rate after 100000 games.

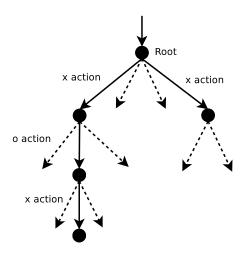


Figure 4: Path finding example, S is the start position, G is the goal position, * are the path points, . are empty spaces and # are walls.