

2020 Spring AI HW-3, Professor: Minkoo Kim

1. A* algorithm를 여러분의 말로 설명하는데 반드시 evaluation function을 언급하여 정확하게 설명하시오. 그리고 A* algorithm을 이용하여 다음의 8-puzzle problem을 해결하는데 매 단계마다 evaluation function이 어떻게 되어 다음 단계로 옮겨지는 정확하게 설명하시오. (10 Points)

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

<start state>

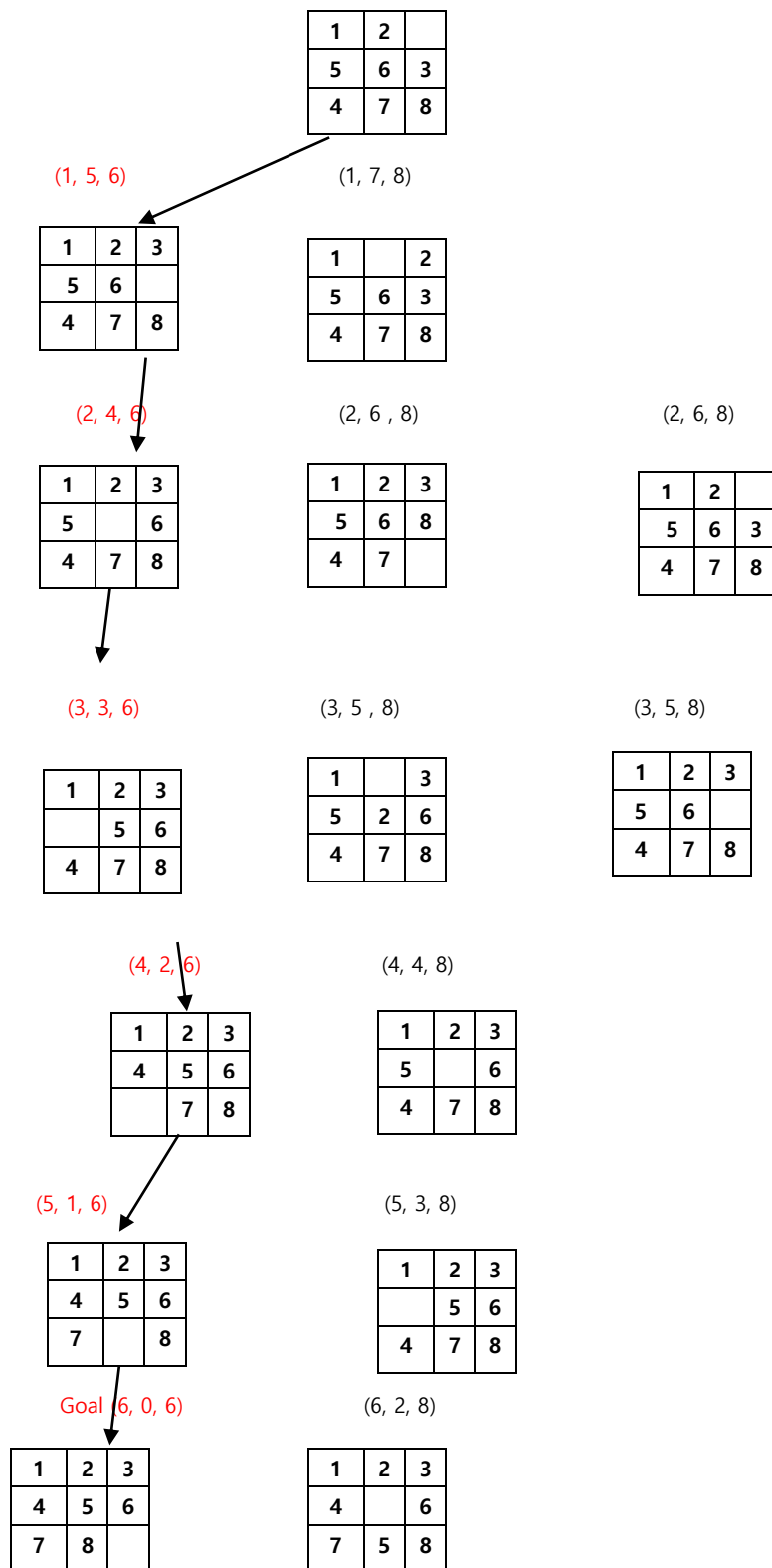
| | | |
|---|---|---|
| 1 | 2 | 3 |
| 4 | 5 | 6 |
| 7 | 8 | |

<goal state>

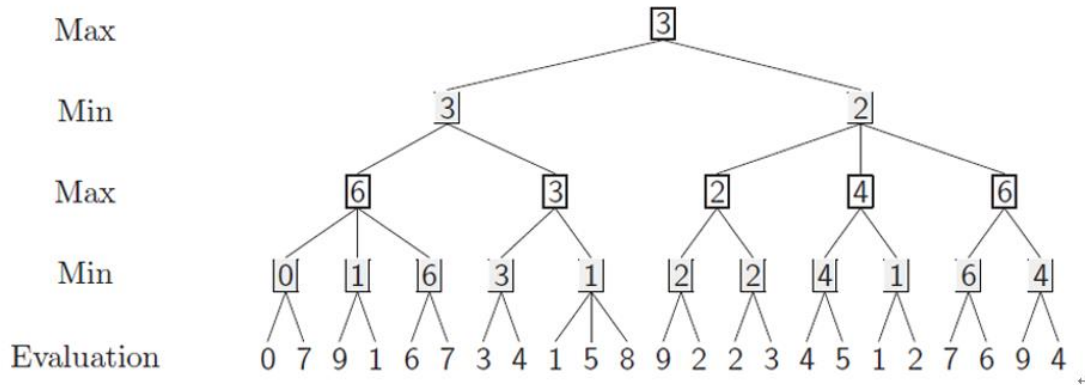
(Solution)

A* search is the best-first search with the following heuristic evaluation function $f(n)$: $f(n) = g(n) + h(n)$ where $g(n)$ is the actual cost from the start node to the current node n and $h(n)$ is the estimated cost from the current node n to the goal node. If f is admissible (that is h is not overestimated), we can find an optimal solution using the A* search algorithm.

I would like to use Manhattan distance as a heuristic function h to solve the above problem. The values in the below () are $g(n)$, $h(n)$, and $f(n)$, respectively.



2. 강의노트에 있는 α - β pruning algorithm을 여러분의 말로 설명하고 다음 문제를 구체적으로 알고리즘을 따라 가면서 α , β 값이 어떻게 변하여 알고리즘이 작동되는지 보이시오. (10 Points)



(Solution)

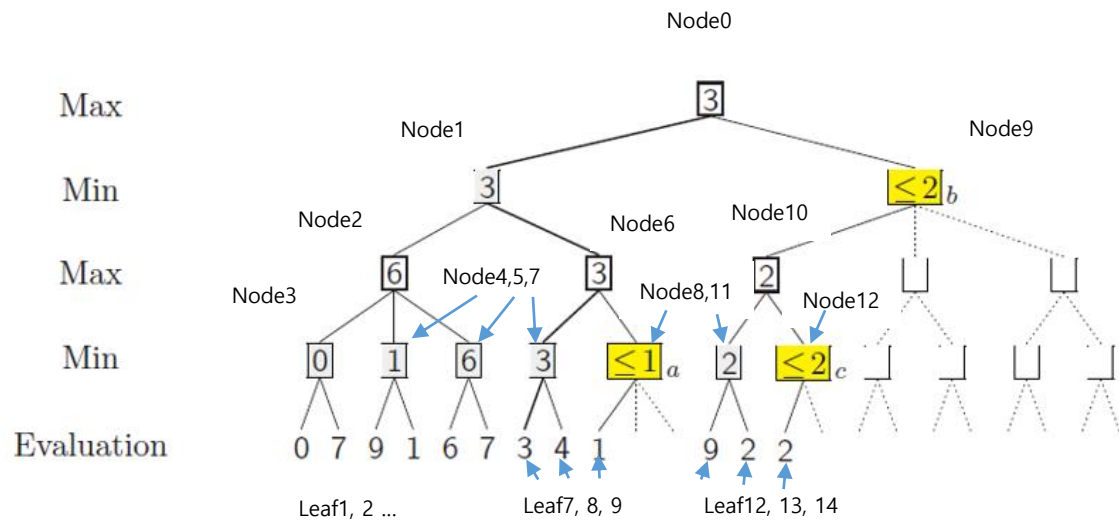
```

AlphaBetaMax(Node,  $\alpha$ ,  $\beta$ )
If DepthLimitReached(Node) Return(Rating(Node))
NewNodes = Successors(Node)
While NewNodes  $\neq \emptyset$ 
     $\alpha$  = Maximum( $\alpha$ , AlphaBetaMin(First(NewNodes),  $\alpha$ ,  $\beta$ ))
    If  $\alpha \geq \beta$  Return( $\beta$ )
    NewNodes = Rest(NewNodes)
Return( $\alpha$ )
    
```

```

AlphaBetaMin(Node,  $\alpha$ ,  $\beta$ )
If DepthLimitReached(Node) Return(Rating(Node))
NewNodes = Successors(Node)
While NewNodes  $\neq \emptyset$ 
     $\beta$  = Minimum( $\beta$ , AlphaBetaMax(First(NewNodes),  $\alpha$ ,  $\beta$ ))
    If  $\beta \leq \alpha$  Return( $\alpha$ )
    NewNodes = Rest(NewNodes)
Return( $\beta$ )
    
```

말로 설명하는 것은 생략되었음.



The nodes are numbered in depth-first manner and leaf nodes are numbered from left to the right.

```

AlphaBetMax(Node0,  $-\infty$ ,  $\infty$ )
  AlphaBetMin(Node1,  $-\infty$ ,  $\infty$ )
    AlphaBetMax(Node2,  $-\infty$ ,  $\infty$ )
      AlphaBetMin(Node3,  $-\infty$ ,  $\infty$ )
        AlphaBetMax(Leaf1,  $-\infty$ ,  $\infty$ )
          Return(0)
         $\beta = 0$ 
        AlphaBetMax(Leaf2,  $-\infty$ , 0)
          Return(7)
        Return(0)
         $\alpha = 0$ 
        AlphaBetMin(Node4, 0,  $\infty$ )
          ...
          Return(1)
         $\alpha = 1$ 
        AlphaBetMin(Node5, 1,  $\infty$ )
          ...
          Return(6)
        Return(6)
         $\beta = 6$ 
        AlphaBetMax(Node6,  $-\infty$ , 6)

```

```

AlphaBetMin(Node7,  $-\infty$ , 6)
  AlphaBetMax(Leaf7,  $-\infty$ , 6)
    Return(3)
   $\beta = 3$ 
  AlphaBetMax(Leaf8,  $-\infty$ , 3)
    Return(4)
  Return(3)
 $\alpha = 3$ 
AlphaBetMin(Node8, 3, 6)
  AlphaBetMax(Leaf9, 3, 6)
    Return(1)
   $\beta = 1$ ; if  $\beta \leq \alpha$  then
    Return(3)
  Return(3)
 $\beta = 3$ 
  Return(3)
 $\alpha = 3$ 
AlphaBetMin(Node9, 3,  $\infty$ )
  AlphaBetMax(Node10, 3,  $\infty$ )
    AlphaBetMin(Node11, 3,  $\infty$ )
      AlphaBetMax(Leaf12, 3,  $\infty$ )
        Return(9)
      AlphaBetMax(Leaf13, 3,  $\infty$ )
        Return(2)
      Return(2)
     $\alpha = 2$ 
    AlphaBetMin(Node12, 2, 3)
      AlphaBetMax(Leaf14, 2, 3)
        Return(2)
       $\beta = 2$ ; if  $\beta \leq \alpha$  then
        Return(2)
     $\beta = 2$ 

```

```
    if  $\beta \leq \alpha$  then Return(2)  
Return(3); /*Max{3, 2} = 3*/
```