

15-Puzzle Problem

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Overview

1 Search

2 IDA*

3 Tasks



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A Search Problem

To formulate a problem as a search problem we need the following components:

- Formulate a **state space** over which to search. The state space necessarily involves abstracting the real problem.
- Formulate **actions** that allow one to move between different states. The actions are abstractions of actions you could actually perform.
- Identify the **initial state** that best represents your current state
- Identify the **goal** or **desired condition** one wants to achieve.
- Formulate various **heuristics** to help guide the search process.



The formalism

- Once the problem has been formulated as a state space search, various algorithms can be utilized to solve the problem.
- A solution to the problem will be a sequence of actions/moves that can transform your current state into state where your desired condition holds.



Algorithms for search

Inputs:

- a specified initial state (a specific world state or a set of world states representing the agent's knowledge, etc.)
- a successor function $S(x)$ = set of states that can be reached from state x via a single action.
- a goal test a function that can be applied to a state and returns true if the state satisfies the goal condition.
- A step cost function $C(x,a,y)$ which determines the cost of moving from state x to state y using action a . ($C(x,a,y) = \infty$ if a does not yield y from x)

Output:

- a sequence of states leading from the initial state to a state satisfying the goal test.



Uninformed search strategies

- These are strategies that adopt a fixed rule for selecting the next state to be expanded.
- The rule does not change irrespective of the search problem being solved.
- These strategies do not take into account any domain specific information about the particular search problem.



Popular uninformed search techniques

- Breadth-First
- Uniform-Cost
- Depth-First
- Depth-Limited
- **Iterative-Deepening search**



Iterative deepening search

- Solve the problems of depth-first and breadth-first by extending depth limited search
- Starting at depth limit $L = 0$, we iteratively increase the depth limit, performing a depth limited search for each depth limit
- Stop if a solution is found, or if the depth limited search failed without cutting off any nodes because of the depth limit
- If no nodes were cut off, the search examined all paths in the state space and found no solution, hence no solution exists.



Iterative deepening search (cont'd)

Iterative deepening search

```
function ITERATIVE-DEEPENING-SEARCH(problem) returns a solution, or failure  
  for depth = 0 to  $\infty$  do  
    result  $\leftarrow$  DEPTH-LIMITED-SEARCH(problem, depth)  
    if result  $\neq$  cutoff then return result
```

- The iterative deepening search algorithm, which repeatedly applies depth limited search with increasing limits.
- It terminates when a solution is found or if the depth limited search returns failure, meaning that no solution exists.



Iterative deepening search (cont'd)

Limit = 0

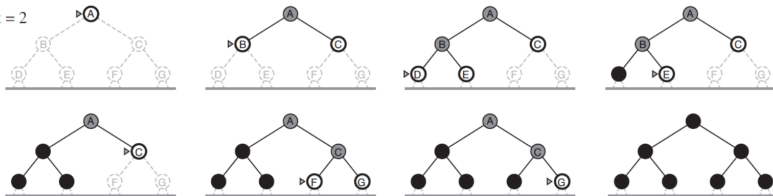


Limit = 1



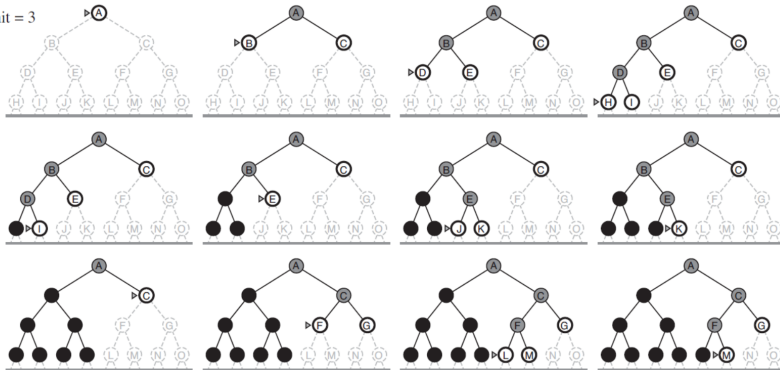
Iterative deepening search (cont'd)

Limit = 2



Iterative deepening search (cont'd)

Limit = 3



Heuristic search

- The idea is to develop a domain specific heuristic function $h(n)$, guessing the cost of getting to the goal from node n
- We require that $h(n) = 0$ for every node n whose state satisfies the goal
- There are different ways of guessing this cost in different domains. i.e., heuristics are domains specific.



- Define an evaluation function $f(n) = g(n) + h(n)$
 - $g(n)$ is the cost of the path to node n
 - $h(n)$ is the heuristic estimate of the cost of getting to a goal node from n
- So $f(n)$ is an estimate of the cost of getting to the goal via node n .
- We use $f(n)$ to order the nodes on the frontier.



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- A* has the same potential space problems as BFS or UCS
- IDA* - Iterative Deepening A* is similar to Iterative Deepening Search and similarly addresses space issues.
- Like iterative deepening, but now the cutoff is the f-value ($g+h$) rather than the depth
- At each iteration, the cutoff value is the smallest f-value of any node that exceeded the cutoff on the previous iteration



Pseudocode

```
1 path                //current search path (acts like a stack)
2 node                //current node (last node in current path)
3 g                  //the cost to reach current node
4 f                  //estimated cost of the cheapest path (root.. node.. goal)
5 h(node)            //estimated cost of the cheapest path (node.. goal)
6 cost(node, succ)   //step cost function
7 is_goal(node)      //goal test
8 successors(node)   //node expanding function, expand nodes ordered by g + h(node)
9 ida_star(root)     //return either NOT_FOUND
10 | | | | |         //or a pair with the best path and its cost
11
12 procedure ida_star(root)
13     bound := h(root)
14     path := [root]
15     loop
16         t := search(path, 0, bound)
17         if t = FOUND then
18             return (path, bound)
19         if t =  $\infty$  then
20             return NOT_FOUND
21         bound := t
22     end loop
23 end procedure
```



Pseudocode (cont'd)

```
1 function search(path, g, bound)
2   node := path.last
3   f := g + h(node)
4   if f > bound then
5     return f
6   if is_goal(node) then
7     return FOUND
8   min := ∞
9   for succ in successors(node) do
10    if succ not in path then
11      path.push(succ)
12      t := search(path, g + cost(node, succ), bound)
13      if t = FOUND then
14        return FOUND
15      if t < min then
16        min := t
17      path.pop()
18    end if
19  end for
20  return min
21 end function
```



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- Please solve 15-Puzzle problem by using IDA* (MATLAB).
- You can use one of the two commonly used heuristic functions:
 - $h1$ = the number of misplaced tiles.
 - $h2$ = the sum of the distances of the tiles from their goal positions.
- Here are 4 test cases for you to verify your algorithm correctness.



Tasks (cont'd)

11	3	1	7
4	6	8	2
15	9	10	13
14	12	5	

TextOut of Result	
11	3 1 7
4	6 8 2
15	9 10 13
14	12 5 0
LowerBound 38 moves	
A optimal solution 56 moves	
Used time 3 sec	
13	10 8 6 9 12 5 13 10 8
12	15 14 5 13 12 15 14 5 13
14	9 4 11 3 1 6 4 11 3
1	6 4 2 8 10 12 15 10 8
7	4 2 11 3 5 9 10 11 3
6	2 3 7 8 12

	5	15	14
7	9	6	13
1	2	12	10
8	11	4	3

TextOut of Result	
0	5 15 14
7	9 6 13
1	2 12 10
8	11 4 3
LowerBound 44 moves	
A optimal solution 62 moves	
Used time 4 sec	
7	9 2 1 8 2 5 7 2 5
1	11 8 9 5 1 6 12 10 3
4	8 11 10 12 13 3 4 8 12
13	15 14 3 4 8 12 13 15 14
7	2 1 5 10 11 13 15 14 7
3	4 8 12 15 14 11 10 9 13
14	15

14	10	6	
4	9	1	8
2	3	5	11
12	13	7	15

TextOut of Result	
14	10 6 0
4	9 1 8
2	3 5 11
12	13 7 15
LowerBound 37 moves	
A optimal solution 49 moves	
Used time 0 sec	
6	10 9 4 14 9 4 1 10 4
1	3 2 14 9 1 3 2 5 11
8	6 4 3 2 5 13 12 14 13
12	7 11 12 7 14 13 9 5 10
6	8 12 7 10 6 7 11 15

6	10	3	15
14	8	7	11
5	1		2
13	12	9	4

TextOut of Result	
6	10 3 15
14	8 7 11
5	1 0 2
13	12 9 4
LowerBound 32 moves	
A optimal solution 48 moves	
Used time 0 sec	
9	12 13 5 1 9 7 11 2 4
12	13 9 7 11 2 15 3 2 15
4	11 15 0 14 1 5 9 13 15
7	14 10 6 1 5 9 13 14 10
6	2 3 4 8 7 11 12



- This program uses IDA* as the basic search algorithm, and this uses WD(Walking Distance) to improve the efficiency of the search.
- WD is a sophisticated lower bound for how many moves are needed to solve an arbitrary board configuration.
- WD gives severe distance than MD(Manhattan Distance).
- As for the details of WD, please read <http://www.ic-net.or.jp/home/takaken/nt/slide/solve15.html>



- Stuart Russell and Peter Norvig, Artificial Intelligence: A Modern Approach.
- IDA*, Wikipedia.
- "15puzzle Optimal solver" for windows, <http://www.ic-net.or.jp/home/takaken/e/15pz/index.html>



The End

