

A* Search

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Objectives

- ► To apply the A* algorithm.
- ► To build the tree of A* search.
- ► To analyse the optimality and complexity of A* search.
- ► To distinguish between A* graph and tree search.



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1 Introduction

 A^* search consist in enumerating paths until it finds a solution, prioritising those of lowest estimated cost (f = g + h) and avoiding cycles:

 A^* generalises UCS with the incorporation of the *heuristic* function h that estimates the minimum cost of reaching a goal state. h is a non-negative function, being zero at a goal state.



2 The A* algorithm (graph search) [1]

```
A* (G, s', h)
                                // G weighed graph, s' start, h heuristic
 O = InitQueue(s', f_{s'} \triangleq 0 + h(s')) // Open: priority queue f \triangleq g + h
  C = \emptyset
                                                // Closed: explored nodes
  while not EmptyQueue(O):
                                             // best-first: s = \arg\min_{n \in O} f_n
                                            // draws in favour of goal state
   s = Pop(O)
   if Goal(s) return s
                                                            // solution found!
   C = C \cup \{s\}
                                                                 // s explored
   for all (s,n) \in Adjacents(G,s):
                                               // generation: n is child of s
     x = (g_s + w(s, n)) + h(n)
                                                           // possibly new f_n
     if n \notin C \cup O: Push(O, n, f_n \triangleq x)
     else if n \in O and x < f_n: Update(O, n, f_n \triangleq x)
     else if n \in C and x < f_n: C = C \setminus \{n\}; Push(O, n, f_n \triangleq x)
  return NULL
                                                        // solution not found
```

3 A* search space

It depens on h(n), minimum cost estimate from n to goal state, $h^*(n)$:



4 Optimality and complexity [1, 2, 3, 4, 5, 6]

- ► Completeness: A* always finishes in finite graphs.
- ► Optimality: If h is admissible, A* returns the optimal solution; it is said that A* is admissible.
- ▶ If h is consistent, nodes are selected to be expanded in non-decreasing order of f, traversing optimum paths $(g = g^*)$.
 - ▷ A* does not re-expand closed nodes (no need to implement it)
 - ▷ A* is equivalent to Dijkstra with *reduced costs* [6].
 - \triangleright A* is *optimally efficient* for h, that is, there is no other algorithm that with the same h would expand fewer nodes [6].
- Complexity: The same as Dijkstra if h is consistent [6].



5 The A* algorithm (tree search)

```
A* (G, s', h)
                               // G weighed graph, s' start, h heuristic
 O = InitQueue(s', f_{s'} \triangleq 0 + h(s')) // Open: priority queue f \triangleq g + h
 while not EmptyQueue(O):
                                          // best-first: s = \arg\min_{n \in O} f_n
                                          // draws in favour of goal state
   s = Pop(O)
   if Goal(s) return s
                                                          // solution found!
   for all (s,n) \in Adjacents(G,s):
                                        // generation: n is child of s
     x = (q_s + w(s, n)) + h(n)
                                                         // possibly new f_n
     if n \notin O: Push(O, n, f_n \triangleq x)
     else if n \in O and x < f_n: Update(O, n, f_n \triangleq x)
  return NULL
                                                      // solution not found
```

6 The A* algorithm (tree search)

 A^* with tree search [5] has no closed nodes ($C = \emptyset$) and re-expands them as they were new nodes:



7 Conclusions

We have studied:

- ► The A* algorithm (graph search).
- ► The A* search space.
- Optimality and complexity in A* search.
- ► The A* algorithm (tree search).

Some aspects to highlight on A*:

- Complete and optimum with positive-cost edges and h admissible.
- ► Simpler and efficient if *h* consistent (closed nodes not re-expanded).
- Excessive spatial cost, especially with deep solutions.
- Reduced spatial cost with tree search.



References

- [1] P. E. Hart, N. J. Nilsson, and B. Raphael. A Formal Basis for the Heuristic Determination of Minimum Cost Paths. *IEEE Transactions on Systems Science and Cybernetics*, 1968.
- [2] J. Pearl. *Heuristics: Intelligent Search Strategies for Computer Problem Solving*. Addison-Wesley, 1984.
- [3] R. Dechter and J. Pearl. Generalized Best-First Search Strategies and the Optimality of A*. *Journal of the ACM*, 1985.
- [4] R. C. Holte. Common Misconceptions Concerning Heuristic Search. In *Proc. of SOCS-10*, 2010.
- [5] S. Russell and P. Norvig. Artificial Intelligence: A Modern Approach. Pearson, third edition, 2010.
- [6] S. Edelkamp and S. Schrödl. Heuristic Search Theory and Applications. Academic Press, 2012.



____ astar.py _____

```
#!/usr/bin/env python3
import heapq
G = \{ A' : [(B', 1), (C', 4)], B' : [(A', 1), (D', 1)], \}
 \rightarrow 'C': [('A', 4), ('E', 1)], 'D': [('B', 1), ('E', 4)],
\rightarrow 'E': [('C',1),('D',4)]}
h0={ 'A':0, 'B':0, 'C':0, 'D':0, 'E':0}
hstar={'A':5,'B':5,'C':1,'D':4,'E':0}
def astar(G,s,t,h):
 \rightarrowOd={s:0}; Cd={} # Open and Closed g dict
  \rightarrowOh=[]; heapq.heappush(Oh,(h[s],s,[s])) # Open heap
 \rightarrowwhile Od:
  \rightarrow \rightarrow s=None
   \rightarrow \rightarrow while s not in Od: fs,s,path=heapq.heappop(Oh) # delete-min

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    \rightarrow \rightarrow if s==t: return qs, path
    \rightarrow \rightarrow del Od[s]; Cd[s]=qs

ightarrowfor n,wsn in G[s]:
       \rightarrow \rightarrow \rightarrow qn=qs+wsn

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    \rightarrow \rightarrow \rightarrowelif n in Od and gn>=Od[n]: continue
   \rightarrow \rightarrow \rightarrow Od[n] = gn; heapq.heappush(Oh, (gn+h[n], n, path+[n]))
print(astar(G, 'A', 'E', h0))
print(astar(G, 'A', 'E', hstar))
                                                                                                                                                                                                        ____ astar.py.out __
```

```
(5, ['A', 'C', 'E'])
(5, ['A', 'C', 'E'])
```

```
___ astar_pqdict.py ____
#!/usr/bin/env python3
from pqdict import pqdict
G = \{ 'A' : [ ('B', 1), ('C', 4)], 'B' : [ ('A', 1), ('D', 1)], 
   'C': [('A', 4), ('E', 1)], 'D': [('B', 1), ('E', 4)],
   'E': [('C',1),('D',4)]}
h0={ 'A':0, 'B':0, 'C':0, 'D':0, 'E':0}
hstar={'A':5,'B':5,'C':1,'D':4,'E':0}
def astar(G,s,t,h):
  O=pqdict(\{s:(0,h[s],[s])\}, key=lambda x:x[0]+x[1]); C=\{\}
  while O:
    s, (qs, hs, path) = 0.popitem()
    if s==t: return qs, path
    C[s]=qs,hs
    for n, wsn in G[s]:
      qn=qs+wsn
      if n in C:
         if qn>=C[n][0]: continue
        oqn, ohn=C[n]; del C[n]; O[n]=qn, ohn, path+[n]
      elif n in O:
         if qn \ge 0[n][0]: continue
         ogn, ohn, opath=O[n]; O[n]=qn, ohn, path+[n]
      else: O[n]=qn,h[n],path+[n]
print(astar(G, 'A', 'E', h0))
```

```
astar_pqdict.py.out ______
(5, ['A', 'C', 'E'])
(5, ['A', 'C', 'E'])
```

print(astar(G, 'A', 'E', hstar))