

Overview

- written as A* Algorithm
- [[Shortest Path Algorithms]] to find a single destination
- based on [[Breadth-First Search]] and [[Dijkstra's Algorithm]]
- informed
 - does not search uniformly
 - uses heuristics
 - prioritizes towards the direction of the goal

Heuristics

Def: A heuristic is **consistent** if for every edge $\{u, v\} \in E$ we have
$$h(u) \leq w(u, v) + h(v)$$

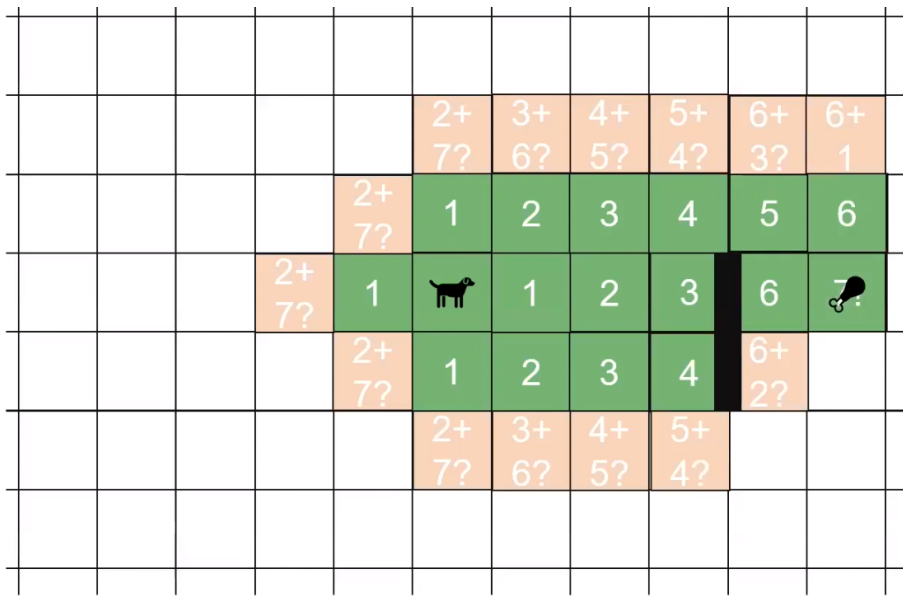
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- perfect heuristic
 - border line impossible
 - requires perfect knowledge lol
- overestimate
 - fast
 - not admissible
 - might not find path even if one exists

A* Heuristics

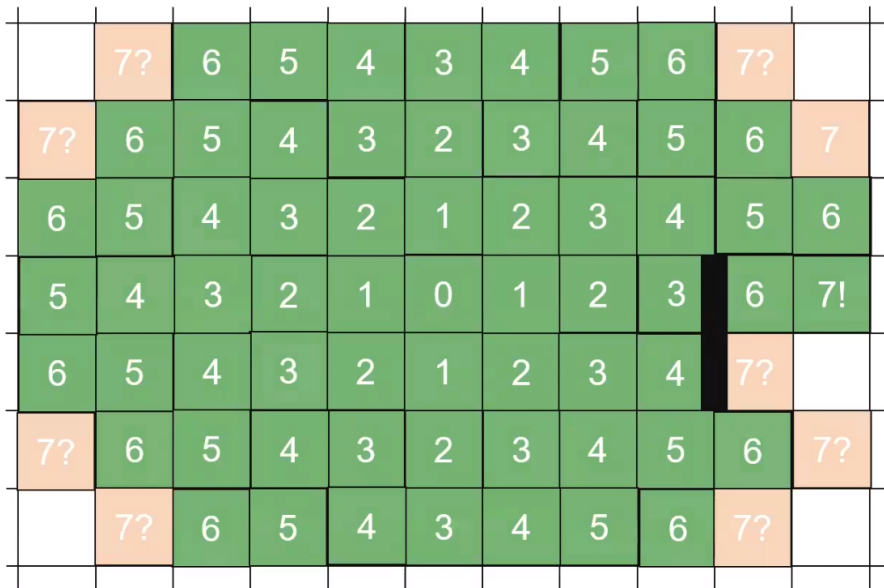
- $g(v)$
 - distance from start to current vertex
- $h(u)$
 - distance from current vertex to end
 - * “Luftlinie” - as the crow flies
 - ignores obstacles which may block the path
 - underestimates the future cost
 - * good characteristic
- therefore pretty efficient

Comparison between A* and Dijkstra's

- A*



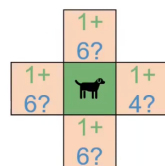
- Dijkstra's



Algorithm

Input: $G(V, E, W)$ start point: s , end point: t

Initialize $S = \{s\}$, $g(s) = 0$, $g(v) = \infty$, $g(v) = w(s, v)$, $v \in N(s)$



S : expanded/closed vertices
 $V \setminus S$: the open vertices

s

$g(v) + h(u)?$

$g(v)$ is the length of the best known (!) path from s to v

While $t \notin S$ do

→ $u = \operatorname{argmin}_{u \in V \setminus S} \{g(u) + h(u, t)\}$

$g(v) + h(u)?$

$g(v)$ is the length of the best known (!) path from s to v

For v s.t. $\{u, v\} \in E$ do

$\text{temp} = \min\{g(v), g(u) + w(u, v)\}$

If $\text{temp} < g(v)$ then:

$g(v) = \text{temp}$

$h(u, t)$ is a heuristic guess for the path from u to t .

→ $S = S \setminus \{v\}$ //does nothing if $v \notin S$

$S = S \cup \{u\}$

S might decrease!

•

– red parts differ from [[Dijkstra's Algorithm]]

Properties

- nodes may expand more than once
 - $g(v)$ heuristic value can change
 - always terminates if a path exists
- optimal if a path exists

Lemma 1: Always, for every open node v and every optimal path P from s to v , there exists an **open node** u on P with $g(u) = \text{distance}(s, u)$

Proof: Let $P = (s = v_0, v_1, v_2, \dots, v = v_k)$.



$C = \{v_i \in P \mid v_i \text{ closed}, g(v_i) = d(s, u)\} \neq \emptyset$,

- Let v^* , the vertex in C with highest index. $v^* \neq v$.
- Let u be the successor of v^* in P (possibly $u = v$).

P is optimal path

$$g(u) \leq g(v^*) + w(v^*, u) = d(s, v^*) + w(v^*, u) = d(s, u) \leq g(u)$$

v^* expanded

definition of v^*

always

u has to be open by definition of C

Corollary: Suppose h is admissible and A^* has not terminated. Then, for any optimal path P from s to t , there is an open node u with

$$g(u) + \text{heuristic}(u, t) \leq \text{distance}(s, t)$$

Proof: By Lemma 1, we have open node $u \in P$ with $g(u) = \text{distance}(s, u)$

$$g(u) + \text{heuristic}(u, t) = \text{distance}(s, u) + \text{heuristic}(u, t)$$

h admissible

$$\leq \text{distance}(s, u) + \text{futureCost}(u, t)$$

Proof (Optimality):

Suppose A^* terminates at t with a suboptimal path, i.e., in the last step we expanded t with

$$g(t) + \overset{0}{\text{heuristic}(t, t)} > \text{distance}(s, t)$$

But, by the corollary, there existed just before “expanding t ”, there is an open node u on an optimal path with

$$g(u) + \text{heuristic}(u, t) \leq \text{distance}(s, t)$$

Contradiction!

- optimally efficient
 - with regards to the number of vertices expanded
- space as bottle neck

– $g(v) + h(v)$ is stored for each visited v

15-Puzzle: Search space has a node for each configuration:

– $16! = 20.922.789.888.000$ vertices!

- motivation for memory bounded heuristic search
- * Iterative deepening A*