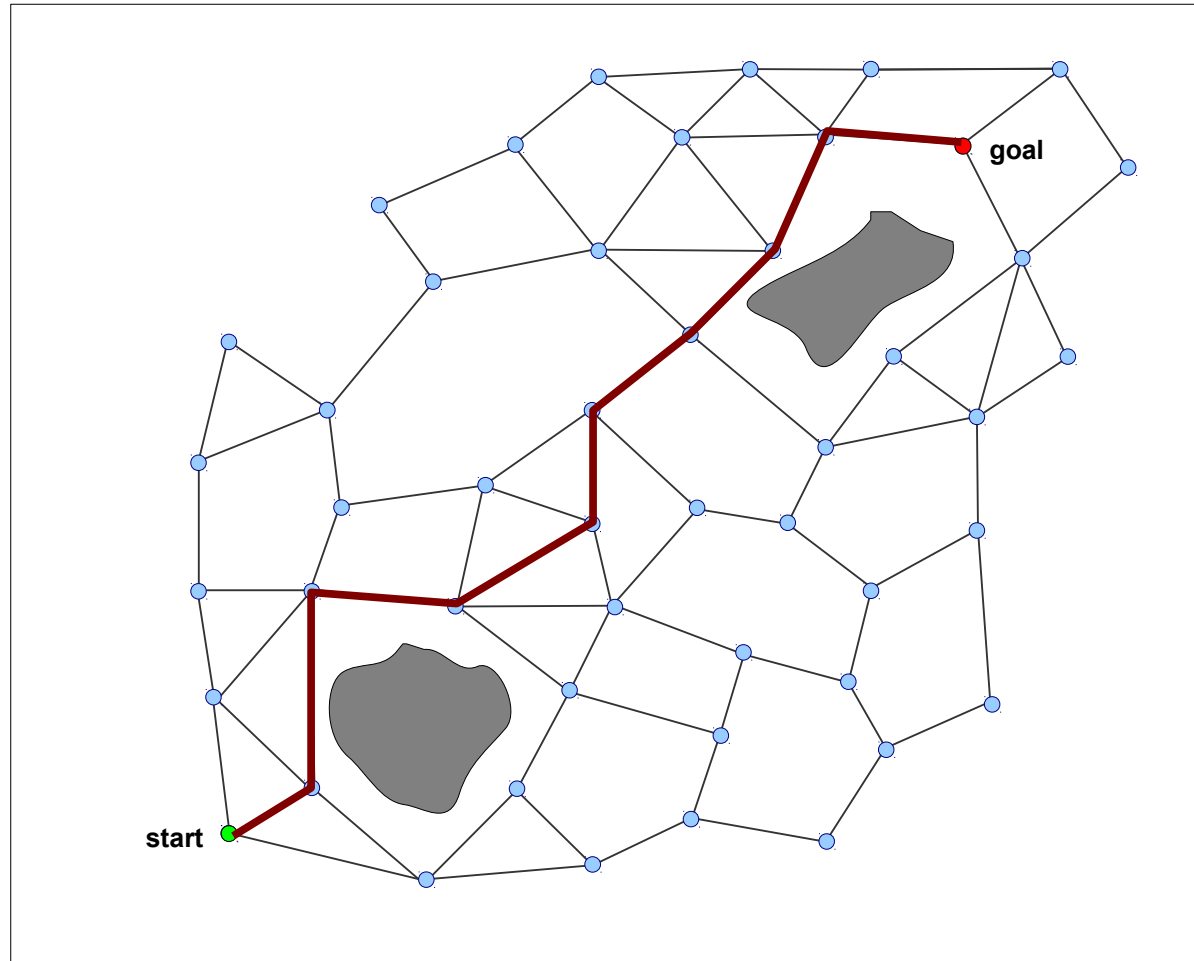


# Introduction to Search-based Planning

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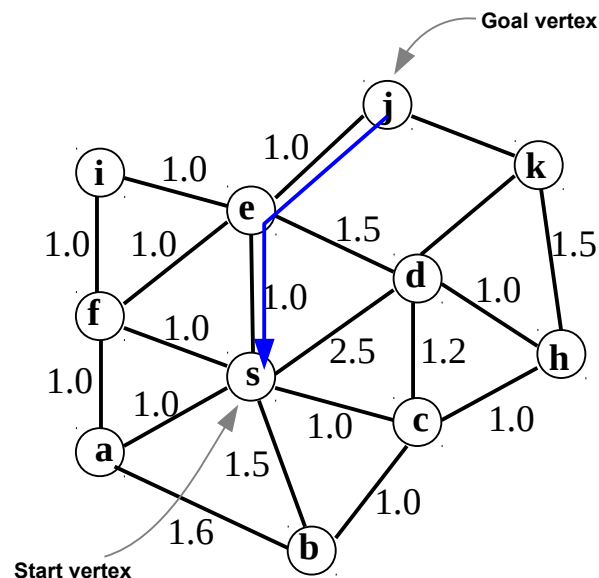
# Search-based Planning

## (Trajectory Planning Using Graph Search)



# Optimal Path Planning in a Graph

## Dijkstra's search algorithm:



Open list

$g$ -scores ( $g: V \rightarrow \mathbb{R}_+$ )

~~$g(s) = 0.0$~~

~~$g(a) = 1.0$~~

~~$g(b) = 1.5$~~

~~$g(c) = 1.0$~~

~~$g(d) = 2.2$~~

~~$g(e) = 1.0$~~

~~$g(f) = 1.0$~~

~~$g(h) = 2.0$~~

~~$g(i) = 2.0$~~

~~$g(j) = 2.0$~~

$g(k) = 3.5$

## Additional observations/details:

- At any instant there are 3 disjoint sets of vertices: **Open list**, **closed list**, and *the rest*.
- **Heap data structure** – efficient maintenance of an ordered list with fast insertion operations (for maintaining  $g$ -scores for the *open list*).
- Worst complexity:  $O(|E| + |V| \log |V|)$

# Pseudocode (Dijkstra's)

$g = \mathbf{Dijkstras}(G, p)$

Inputs:     a. Graph  $G$   
           b. Start node  $p \in \mathcal{V}(G)$

Outputs:    a. The shortest distance map  $g : \mathcal{V}(G) \rightarrow \mathbb{R}^+$

```
1  Initiate  $g$ : Set  $g(v) := \infty$ , for all  $v \in \mathcal{V}(G)$    // Minimum distance
2  Set  $g(p) = 0$ 
3  Set  $Q := \{p\}$    // Open list
4  while ( $Q \neq \emptyset$  AND stopping criterion not satisfied)
5       $q := \operatorname{argmin}_{q' \in Q} g(q')$    // Vertex to expand.  $Q$  is maintained by a heap data-structure.
6      if ( $g(q) == \infty$ )
7          break
8       $Q = Q - q$    // Remove  $q$  from  $Q$ 
9      for each ( $\{w \in \mathcal{N}_G(q) \mid w \in Q \text{ OR } g(w) == \infty\}$ )   // For each unexpanded neighbor of  $q$ 
10         Set  $g' := g(q) + \mathcal{C}_G([q, w])$ 
11         if ( $g' < g(w)$ )
12             Set  $g(w) = g'$ 
13             Set  $Q = Q \cup \{w\}$    // Insert in open list if not already there.
14 return  $g$ 
```

# Pseudocode (Path Reconstruction)

$P = \text{Reconstruct\_Path}(G, g, r)$

Inputs:     a. Graph  $G$   
             b. The shortest distance map  $g : \mathcal{V}(G) \rightarrow \mathbb{R}^+$   
             c. Vertex to which to find the shortest path,  $r \in \mathcal{V}(G)$

Outputs:    a. A path (ordered set of vertices) in the graph,  $P = [\rho_1, \rho_2, \rho_3, \dots, \rho_n = r]$

```
1  Initiate  $P = []$ 
2  if ( $g(r) == \infty$ )     //  $r$  unreachable from the start node
3      return  $P$ 
4  Set  $v := r$ 
5  while ( $g(v) \neq 0$ )
6       $P = v \oplus P$      // Insert  $v$  at the beginning of  $P$ .
7       $v = \operatorname{argmin}_{w \in \mathcal{P}_{G,g}(v)} g(w)$      // back-trace predecessor that led to  $v$ .
8   $P = v \oplus P$      // Insert the final vertex (the start node) at the beginning of  $P$ .
9  return  $P$ 
```

Where,  $\mathcal{P}_{G,g}(u) = \{w' \in \mathcal{V}(G) \mid [w', u] \in \mathcal{E}(G), g(u) = g(w') + \mathcal{C}_G([w', u])\}$   
is the set of potential *predecessors* of  $u$ .

# Pseudocode ( $A^*$ )

$P = A^*(G, p, r, h_r)$

Inputs:     a. Graph  $G$   
              b. Start node  $p \in \mathcal{V}(G)$   
              c. Goal node  $r \in \mathcal{V}(G)$   
              d. An admissible heuristic function,  $h_r : \mathcal{V}(G) \rightarrow \mathbb{R}_+$

Outputs:    a. A path connecting start vertex to goal vertex,  $P = [p = \rho_1, \rho_2, \rho_3, \dots, \rho_n = r]$

```
1  Initiate  $g, f$ : Set  $g(v) := \infty$ ,  $f(v) := \infty$ , for all  $v \in \mathcal{V}(G)$ 
2  Set  $g(p) = 0$  and  $f(p) = h(p)$ 
3  Set  $Q := \{p\}$      // Open list
4  while ( $Q \neq \emptyset$  AND stopping criterion not satisfied)
5       $q := \operatorname{argmin}_{q' \in Q} f(q')$      // Vertex to expand.  $Q$  is maintained by a heap data-structure.
9      if ( $q == r$ )     // Goal vertex reached.
10         return Reconstruct_Path ( $G, g, r$ )
6      if ( $g(q) == \infty$ )
7         break     // No path found.
8       $Q = Q - q$      // Remove  $q$  from  $Q$ 
9      for each ( $\{w \in \mathcal{N}_G(q) \mid w \in Q \text{ OR } g(w) == \infty\}$ )     // For each unexpanded neighbor of  $q$ 
10         Set  $g' := g(q) + \mathcal{C}_G([q, w])$ 
11         if ( $g' < g(w)$ )
12             Set  $g(w) = g'$ 
18             Set  $f(w) = g' + h_r(w)$ 
13             Set  $Q = Q \cup \{w\}$      // Insert in open list if not already there.
14 return []
```

**Admissible heuristics function:**  $h_r(v)$  should be less than or equal to cost of shortest path from  $v$  to  $r$ .

# Properties of $A^*$

- If  $h_r$  is admissible, then  $A^*$  is guaranteed to return the optimal path connecting the start and goal,  $r$ .
- Dijkstra's search is equivalent to  $A^*$  search with zero heuristics.
- **Weighted  $A^*$ :** If the heuristic function used,  $\bar{h}_r$ , is such that
$$\bar{h}_r \leq \varepsilon h_r \quad (\text{where, } h_r \text{ is admissible, } \varepsilon > 1)$$
then, the computed path is at most  $\varepsilon$ -suboptimal (the cost of the returned path is at most  $\varepsilon$  times the cost of the optimal path).

# Optimal Path Planning in a Graph

## Dijkstra's

goal



start

## A\* algorithm

goal

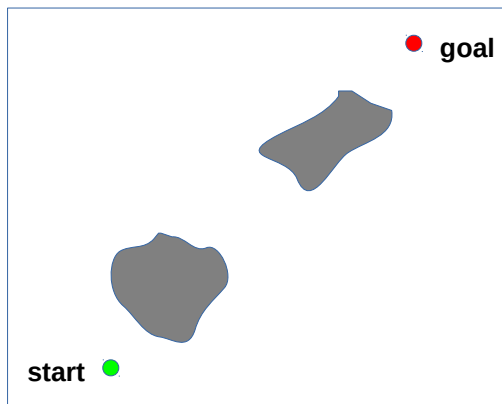


start

Speed up search (expand less vertices) using an ***admissible heuristic function*** ( $h: V \times V \rightarrow \mathbb{R}_+$ ) that underestimates the true least cost between vertices.

*Still guarantees optimality.*

## Features of Dijkstra's and A\*:



- The complete graph need not be available to start with.
- We only need to be able to query the list of neighboring vertices of vertex that we are ***expanding***, and cost of the edges connecting to them.  
(Ex: useful when, for example, the obstacles are given as semi-algebraic sets).
- The graph itself may be infinite/unbounded.



# C++ Library for Graph Search

- DOSL (Discrete Optimal Search Library) :  
Available at <https://github.com/subh83/DOSL>