

Motion planning

Chapter 10
Introduction to Robotics:
Mechanics, Planning, and Control
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Some definitions

- q is the configuration (e.g., $q = (\theta_1, ..., \theta_n)$)
- C is the configuration space (C-space)
- C_{free} is the collision-free C-space
- x = (q, v) is the state
- q(x) returns the configuration associated with x
- *u* is a control
- *U* is the set of possible controls





The problem definition

Equations of motion:

$$\dot{x}=f(x,u)$$
 or $x(T)=x(0)+\int_0^T f(x(t),u(t))dt$ (10.2)

Given an initial state $x(0) = x_{start}$ and a desired final state x_{goal} , find a time T and a set of controls $u : [0,T] \to \mathcal{U}$ such that the motion (10.2) satisfies $x(T) = x_{goal}$ and $q(x(t)) \in \mathcal{C}_{free}$ for all $t \in [0,T]$.





Types of motion planning problems

- Path planning (piano mover's) vs. motion planning
- Fully actuated vs. underactuated (e.g., a car)
- Online vs. offline
- Optimal vs. satisficing
- Exact vs. approximate
- With vs. without obstacles





Properties of motion planners

- Multiple queries vs. single queries
- "Anytime" planning
- Completeness
 - complete
 - resolution complete
 - probabilistically complete
- Computational complexity





Motion planning methods

- Complete methods
- Grid methods
- Sampling methods
- Virtual potential fields
- Nonlinear optimization





A summary of (many) path planning methods

robot + obstacles \Rightarrow discretized graph representation of $C_{\text{free}} \Rightarrow$ search of that graph for a path from q_{start} to q_{goal}





C-space obstacles: 2R arm

Robot's configuration is treated as a point in C-space. Q: How many connected components does C_{free} have?

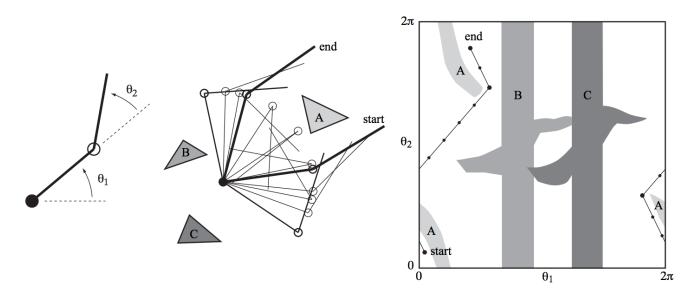


Figure 10.2: (Left) The joint angles of a 2R robot arm. (Middle) The arm navigating among obstacles. (Right) The same motion in C-space.





C-space obstacles: circular mobile robot

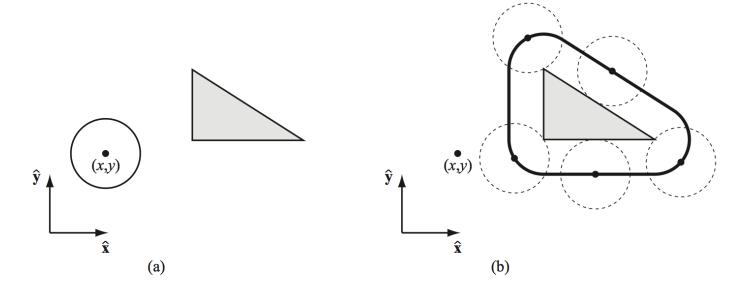
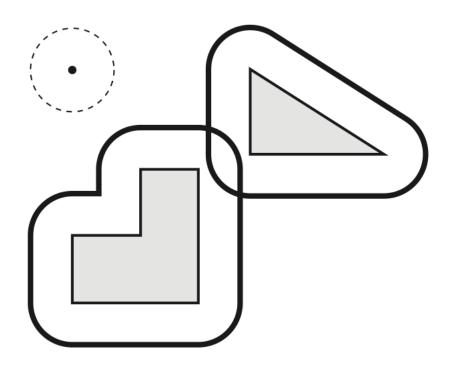


Figure 10.3: (a) A circular mobile robot (white) and a workspace obstacle (grey). The configuration of the robot is represented by (x, y), the center of the robot. (b) In the C-space, the obstacle is "grown" by the radius of the robot and the robot is treated as a point. Any (x, y) configuration outside the dark boundary is collision-free.





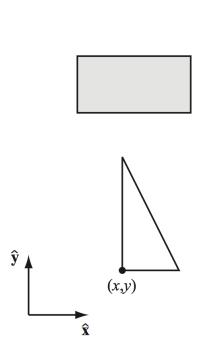
C-space obstacles: circular mobile robot

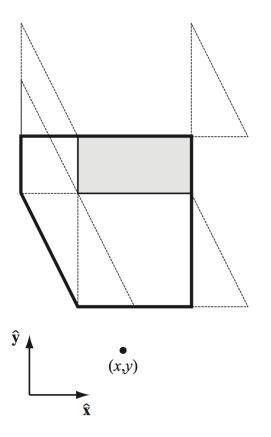






C-space obstacles: polygonal mobile robot

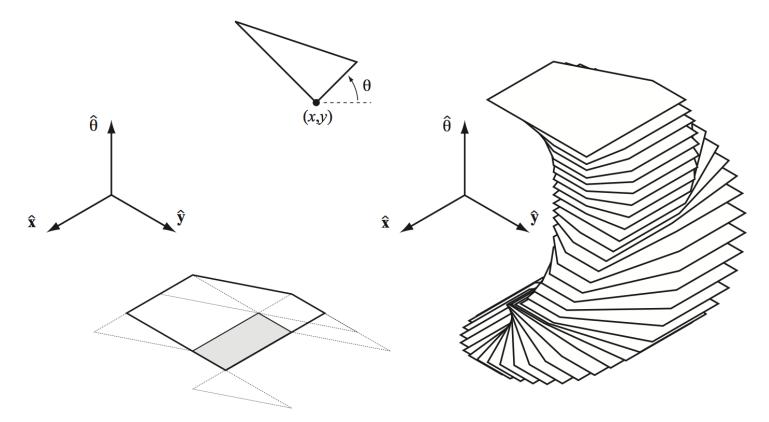








C-space obstacles: rotating/translating polygon





Distance to C-obstacle \mathcal{B}

$$d(q, \mathcal{B}) > 0$$
 no contact with the obstacle

$$d(q, \mathcal{B}) = 0$$
 contact

$$d(q, \mathcal{B}) < 0$$
 penetration.

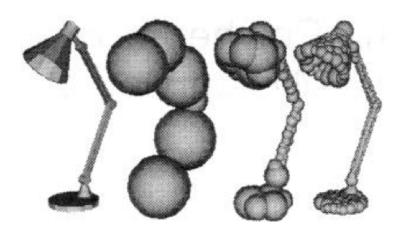
A distance-measurement algorithm returns d.

A *collision-detection algorithm* returns true if q is in collision, false otherwise.





An approximation approach



 $r_i(q)$ is the center of robot sphere i

 R_i is the radius of robot sphere i

 b_i is the center of obstacle sphere j

 \vec{B}_i is the radius of obstacle sphere j

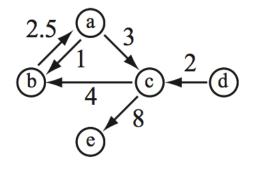
$$d(q, \mathcal{B}) = \min_{i,j} ||r_i(q) - b_j|| - R_i - B_j$$

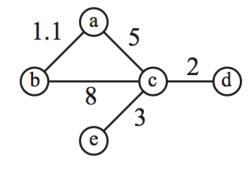
Other primitives (cylinders, rectangular prisms, meshes or polygon soups) can also be used, with increasing accuracy and computational complexity.

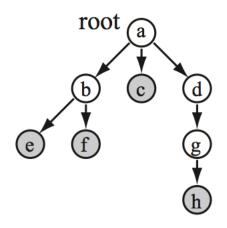




Graphs and trees







weighted directed graph (digraph)

weighted undirected graph

tree (leaves are in gray)





Complete path planning using a roadmap

Reduce the complex high-dimensional $C_{\rm free}$ to a one-dimensional roadmap R (or a graph) with the following properties:

- (i) Reachability. From every point $q \in \mathcal{C}_{\text{free}}$, a free path to a point $q' \in R$ can be found trivially (e.g., a straight-line path).
- (ii) Connectivity. For each connected component of C_{free} , there is one connected component of R.





Constructing a roadmap

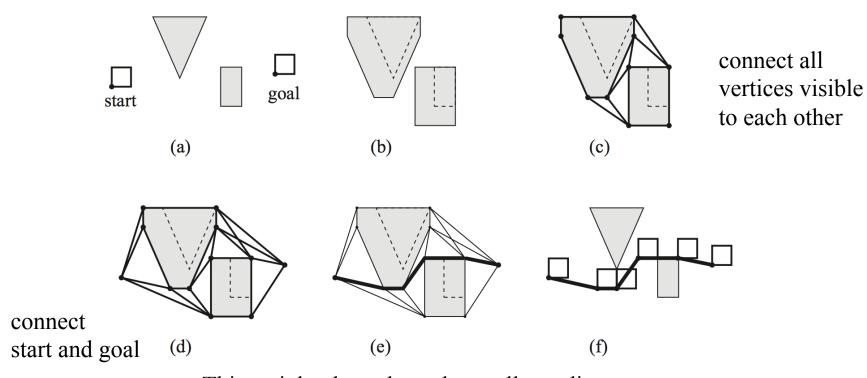
Very difficult to do in general (see John Canny's PhD thesis, 1988), but

- easy for some simple problems
- approximations are possible (e.g., probabilistic roadmap PRM)





Visibility graph for translating polygon

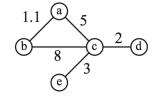




This weighted graph roadmap allows direct search for a shortest path.



Finding a shortest path: A* search



- Keep a list OPEN of nodes reached so far in the search, sorted by the estimated cost of a path going through that node to the goal (the sum of the minimum cost known to reach the node, plus an underestimate to reach the goal).
- Pick the first node from OPEN. If it's at the goal, done.
- For each neighbor of node that has not already been searched from, check if the minimum cost known to reach it (and its parent node) should be replaced.
- Insert these nodes in OPEN.





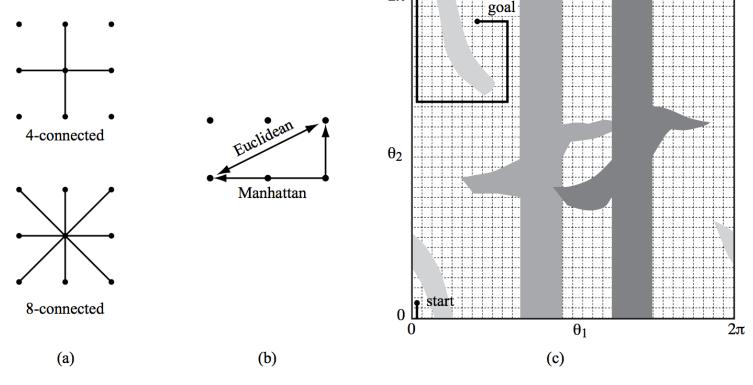
Finding a shortest path: A* search

```
Algorithm 1 A^* search.
1: OPEN \leftarrow \{1\}
2: past_cost[1] \leftarrow 0, past_cost[node] \leftarrow infinity for node \in \{2, \dots, N\}
3: while OPEN is not empty do
     current ← first node in OPEN, remove from OPEN
     add current to CLOSED
     if current is in the goal set then
       return SUCCESS and the path to current
 7:
     end if
     for each nbr of current not in CLOSED do
       tentative_past_cost ← past_cost[current] + cost[current,nbr]
10:
       if tentative_past_cost < past_cost[nbr] then</pre>
11:
         past_cost[nbr] ← tentative_past_cost
12:
         parent[nbr] ← current
13:
         put (or move) nbr in sorted list OPEN according to
14:
               est_total_cost[nbr] ← past_cost[nbr] +
                       heuristic_cost_to_go(nbr)
       end if
15:
     end for
16:
17: end while
18: return FAILURE
```





A* search on a grid





Need not explicitly construct the graph in advance.

