

CS 369: Introduction to Robotics

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Outline for today

- Sampling-based motion planning
- Robot perception

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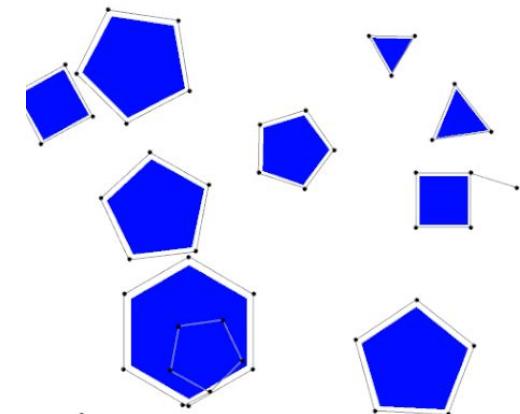
- Sampling-based motion planning
- Robot perception

Motion planning algorithms

	Graph search	Sampling-based
Completeness	Complete	Probabilistically complete
Path optimality	Optimal	Depends
Memory usage	High (store entire map)	Low (store sampled nodes)
Best for	Low-dimensional, 2D grids	High-dimensional, complex spaces

Probabilistic roadmap (PRM)

- Consists of two phases: construction and query
- Construction:
 - Take random samples from the robot's configuration space
 - Connect configurations to neighbors while avoiding collisions with obstacles
 - Add configurations and connections until the roadmap is dense enough
- Query:
 - Connect start and goal configurations to the graph
 - Obtain path with a graph search algorithm



Rapidly exploring random tree (RRT)

Grows a tree rooted at the start configuration by using random samples.

Algorithm BuildRRT

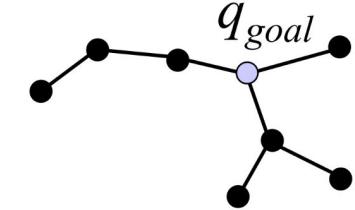
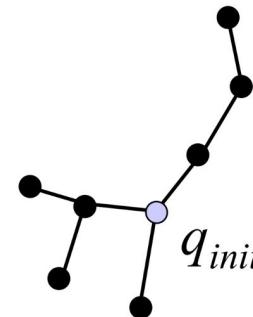
Input: Initial configuration q_{init} , number of vertices in RRT K , incremental distance Δq

Output: RRT tree T

```
T.init( $q_{init}$ )
for  $k = 1$  to  $K$  do
     $q_{rand} \leftarrow \text{RAND\_CONF}()$ 
     $q_{near} \leftarrow \text{NEAREST\_VERTEX}(q_{rand}, T)$ 
     $q_{new} \leftarrow \text{NEW\_CONF}(q_{near}, q_{rand}, \Delta q)$ 
    T.add_vertex( $q_{new}$ )           // Only add if connection is feasible
    T.add_edge( $q_{near}$ ,  $q_{new}$ )
return  $T$ 
```

RRT-Connect

Grows two RRTs towards each other.



```
RRT_CONNECT( $q_{init}, q_{goal}$ )
 $T_a$ .init( $q_{init}$ );  $T_b$ .init( $q_{goal}$ )
for  $k = 1$  to  $K$  do
     $q_{rand} = \text{RANDOM\_CONFIG}()$ 
    if not (EXTEND( $T_a, q_{rand}$ ) = Trapped) then
        if (EXTEND( $T_b, q_{new}$ ) = Reached) then
            return PATH( $T_a, T_b$ )
    SWAP( $T_a, T_b$ ) // Instead of switching, use  $T_a$  as smaller tree
return Failure
```

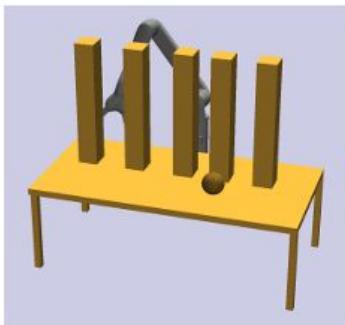
RRT*

- RRT with two improvements:
 - **Best neighbor search:** when adding a new node, RRT* checks nearby nodes in the tree to connect to the one that results in the lowest cost from the root, rather than just the single nearest node
 - **Tree rewiring:** after adding a new node, the algorithm checks if neighboring nodes can have their path cost reduced by switching their parent to the newly added node
- Computationally more expensive, asymptotically optimal

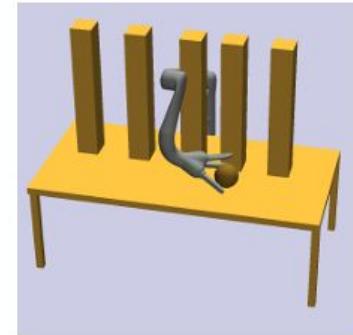
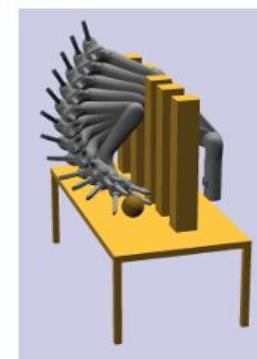
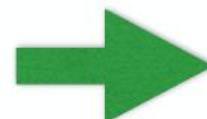
RRT* pseudo code

```
T.init( $q_{init}$ )
for  $i = 0$  to  $max\_iter$ :
     $q_{rand} = \text{RAND\_CONF}()$ 
    if  $\text{OBS}(q_{rand}) == \text{True}$ , try again
     $q_{near} = \text{NEAREST}(q_{rand}, T)$ 
    if not  $\text{EXTEND}(q_{near}, q_{rand}) == \text{Trapped}$ :
         $\text{UPDATE}(q_{rand}, q_{near})$ 
         $q_{best}, neighbors = \text{FIND\_NEIGHBORS}(T, q_{rand}, rad)$ 
        if  $q_{best} != q_{near}$  and not  $\text{EXTEND}(q_{best}, q_{rand}) == \text{Trapped}$ :
             $\text{UPDATE}(q_{rand}, q_{best})$ 
        for  $n$  in  $neighbors$ :
            if  $\text{COST}(q_{rand}) + \text{DIST}(q_{rand}, n) < \text{COST}(n)$  and not  $\text{EXTEND}(q_{rand}, n) == \text{Trapped}$ :
                 $\text{UPDATE}(n, q_{rand})$ 
return  $T$ 
```

Putting it all together



start pose



goal