

# RRT - Rapidly-Exploring Random Trees

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# Collision Free Path Planning

# Motivation

- path planning: find a path from location  $A$  to  $B$
- Example for path planning:
  - mobile robot inside a build
  - shall go to location  $XY$
- Example extension for collision free path planning:
  - avoiding walls and not falling stairs

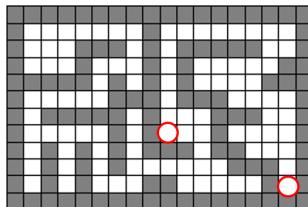


Figure: first example for path planing

# Simple Example

- Simple General Forward Search
  - State: Unvisited, Dead, Alive
  - Priority queue,  $Q$ , with the set of alive states
  - Start loop over  $Q$
  - In each while iteration check next state
    - It is the goal, is terminate
    - Otherwise, it tries applying every possible action

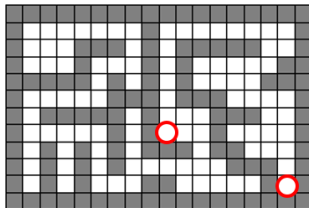


Figure: second example for path planing

# Algorithms

- Other known collision free path planning algorithms
  - Breadth first
  - Deep first
  - Dijkstra's algorithm
  - A\*
  - Best First
  - Backward search
  - ...

# Principles

- Basic Ingredients of Planning
  - State
  - Input
  - Initial and goal states
  - A criterion: Feasibility and/or Optimality
  - a plan

# Rapidly-Exploring Random Trees

# Principles

- Grows a tree rooted at the starting configuration by using random samples from the search space
- As each sample is drawn, a connection is attempted between it and the nearest state in the tree
- If the connection is feasible, this results in the addition of the new state to the tree
- The probability of expanding an existing state is proportional to the size of its Voronoi region
  - As the largest Voronoi regions belong to the states on the frontier of the search = the tree preferentially expands towards large unsearched areas

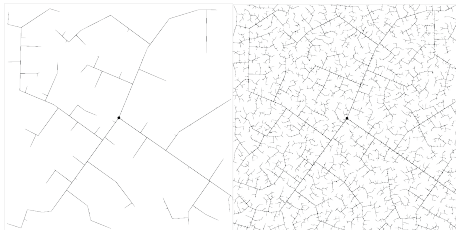


Figure: rrt with 45 and 2345 iterations



# Nice Properties

- The expansion is heavily biased toward unexplored portions of the state space
- The distribution of vertices approaches the sampling distribution, leading to consistent behavior
- Is probabilistically complete under very general conditions
- The algorithm is relatively simple, which facilitates performance analysis

# Nice Properties

- It always remains connected, even though the number of edges is minimal
- Can be considered as a path planning module, which can be adapted and incorporated into a wide variety of planning systems
- Entire path planning algorithms can be constructed without requiring the ability to steer the system between two prescribed states, which greatly broadens the applicability of RRTs

# Notations

$T$  = RRT (Tree of vertices)

$C$  = configuration space of a rigid body or  
systems of bodies in a world

$T(C)$  = target bundle of the configuration space

$C_{goal}$  = goal region,  $C_{goal} \subset C$

$C_{obs}$  = obstacle region,  $C_{obs} \subset C$

$C_{free}$  = region without obstacles,  $C_{free} \subset C$

$q_{init}$  = initale state

$q_n$  = neighbor of a state

$alpha$  = random state

$edges$  = correspond to a path that lies entirely in  $C_{free}$

# Pseudo Code

```
1 generate_rt(robot, vertex_count, delta_time):
2 {
3     q_init = is the current configuration of the robot
4
5     T(q_init)
6
7     for i to vertex_count do
8         alpha = generate_random_state(robot)
9         q_n = find_nearest_neighbor(robot, alpha, T)
10        q_s = generate_state(robot, q_n, alpha, delta_time)
11        T.insert_state(q_s)
12        T.insert_edge(q_n, q_s)
13
14    return T;
15 }
```

Listing 1: pseudocode for rrt algorithm

## Function: generate random state

- generate a random state between the minimum and the maximum configuration limits of the robot

```
1 generate_random_state(robot):  
2 {  
3     min_angle = the minimum angle limits of the robot  
4     max_angle = the minimum angle limits of the robot  
5  
6     return min_angle + (max_angle - min_angle) * random value  
           between 0 and 1  
7 }
```

Listing 2: pseudocode for random state generation

# Function: find nearest neighbor

- bla

## Function: generate state

- checked the path between the  $q_n$  and the  $\alpha$  of collisions free
- if it collisions, then give back the last state before the collision
- if the time for trajectory larger as the  $\delta_{time}$ , then give back the state at  $\delta_{time}$
- else, then give back the  $\alpha$

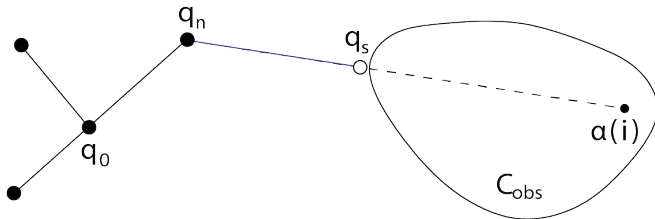


Figure: generate state methode

# Challenges of our work

überlegung 2d -> 3d -> 6d / configuration space



# Live Demo

graph aufbau vollständigen graphen

# Sources

- 1 Rapidly-Exploring Random Trees: A New Tool for Path Planning - Steven M. LaValle  
`http://coitweb.uncc.edu/~xiao/itcs6151-8151/RRT.pdf`  
(03.02.2015)
- 2 Planning Algorithms - Steven M. LaValle  
`http://planning.cs.uiuc.edu/` (03.02.2015)
- 3 `http://en.wikipedia.org/wiki/Rapidly_exploring_random_tree` (03.02.2015)

# Thank you for your attention