

# **Autonomous and Mobile Robotics**

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## **Motion Planning 2**

# **Probabilistic Planning**

DIPARTIMENTO DI INGEGNERIA INFORMATICA  
AUTOMATICA E GESTIONALE ANTONIO RUBERTI



**SAPIENZA**  
UNIVERSITÀ DI ROMA

# sampling-based methods

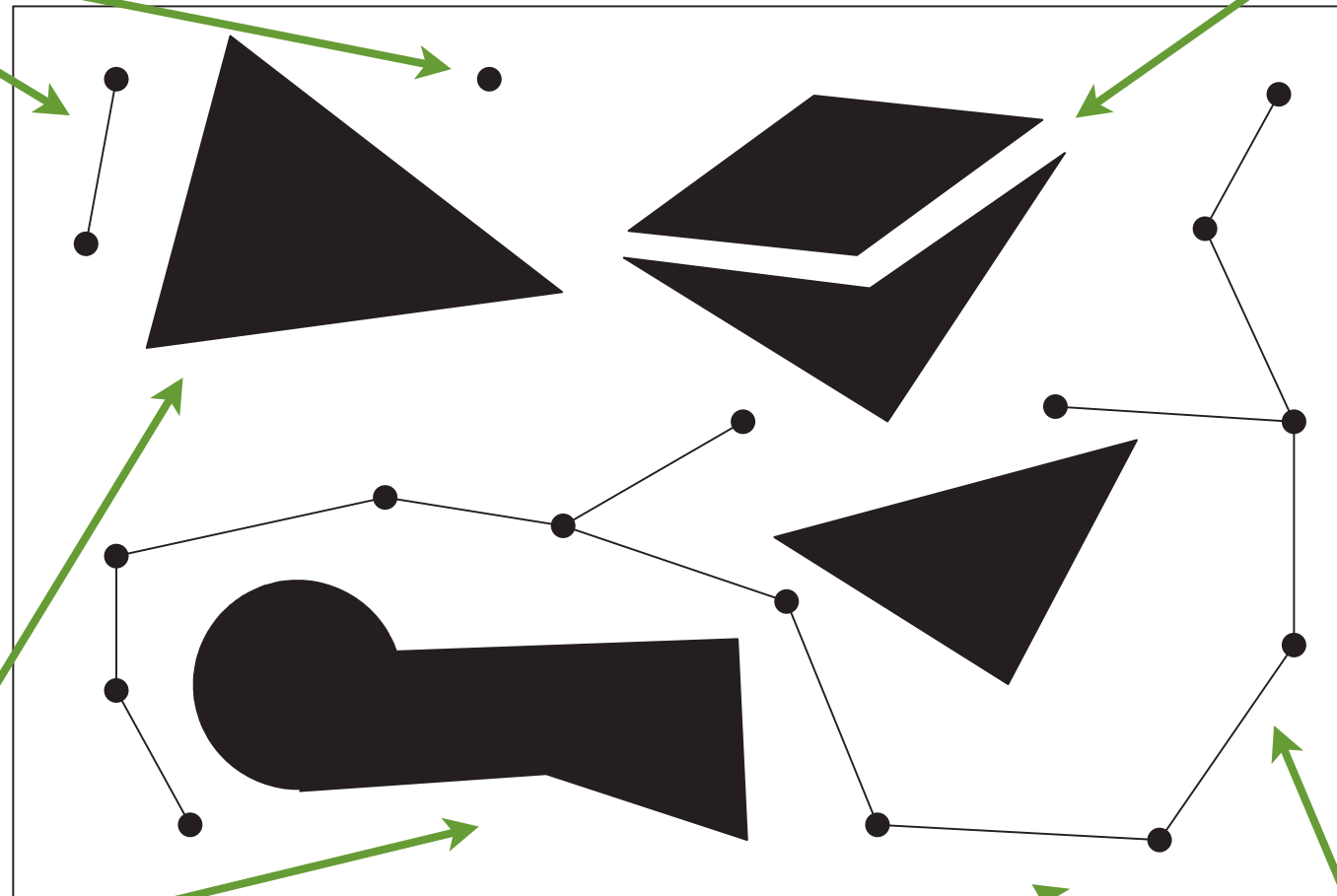
- build a roadmap of the configuration space  $\mathcal{C}$  by repeating this basic iteration:
  - extract a **sample**  $q$  of  $\mathcal{C}$
  - use forward kinematics to compute the **volume**  $\mathcal{B}(q)$  occupied by the robot  $\mathcal{B}$  at  $q$
  - check **collision** between  $\mathcal{B}(q)$  and obstacles  $\mathcal{O}_1, \dots, \mathcal{O}_p$
  - if  $q \in \mathcal{C}_{\text{free}}$ , **add**  $q$  to the roadmap; else, **discard** it
- preliminary computation of  $\mathcal{CO}$  is completely **avoided**: an approximate representation of  $\mathcal{C}_{\text{free}}$  is directly built as a collection of connected configurations (roadmap)
- different criteria for sampling lead to different methods: in general, **randomized outperforms deterministic**

# PRM (Probabilistic Roadmap)

- basic iteration to build the PRM:
  - extract a **sample**  $q$  of  $\mathcal{C}$  with **uniform probability distribution**
  - compute  $\mathcal{B}(q)$  and check for **collision**
  - if  $q \in \mathcal{C}_{\text{free}}$ , **add**  $q$  to the PRM; else, **discard** it
  - search the PRM for “**sufficiently near**” configurations  $q_{\text{near}}$
  - if possible, connect  $q$  to  $q_{\text{near}}$  with a **free local path**
- the generation of a free path between  $q$  and  $q_{\text{near}}$  is delegated to a procedure called **local planner**: e.g., throw a linear path and check it for collision
- the chosen **metric** in  $\mathcal{C}$  plays a role in identifying  $q_{\text{near}}$

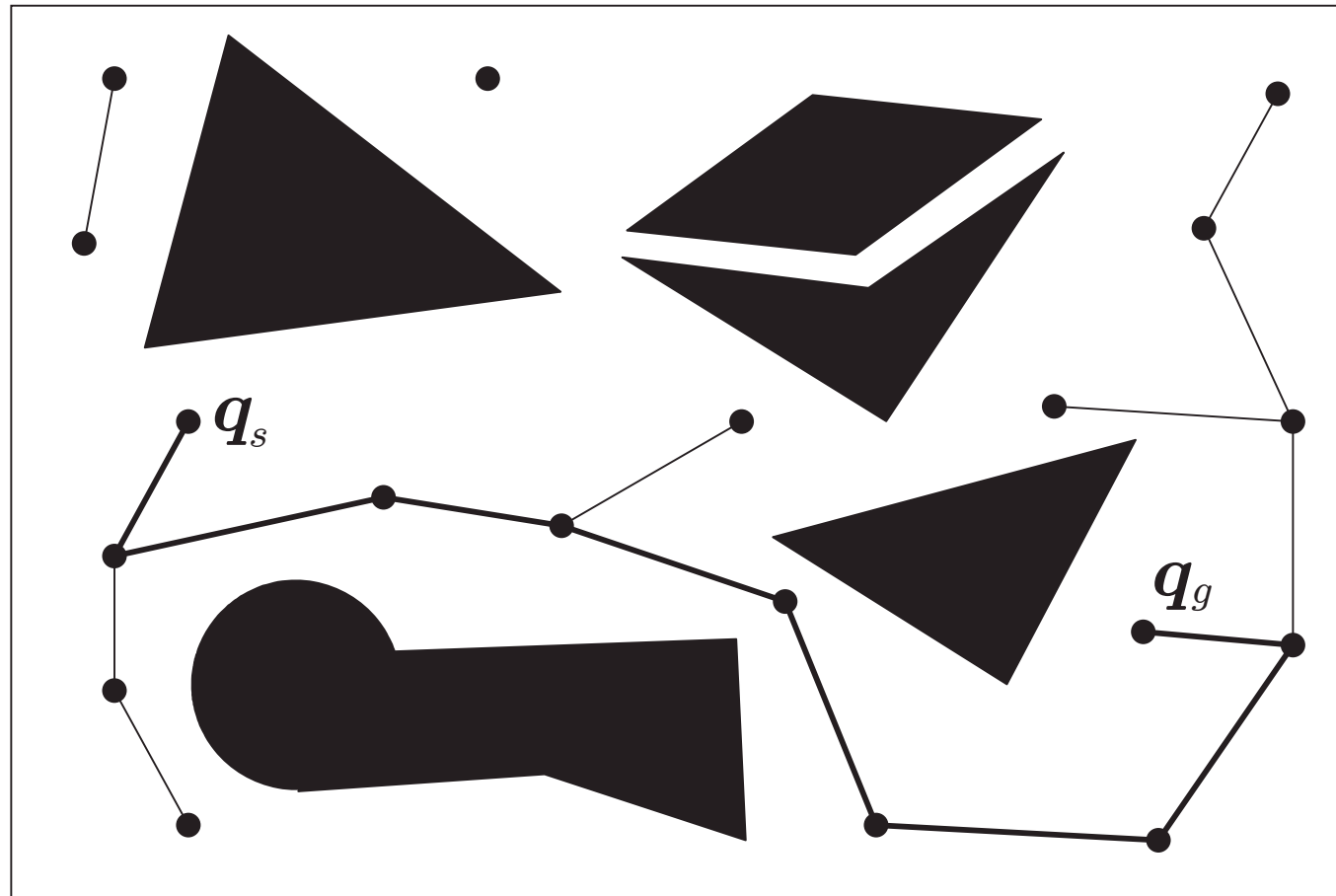
disconnected  
components

narrow passages  
are scarcely sampled



$\mathcal{C}$ -obstacles are  
**never** computed

**local**  
paths



- construction of the PRM is **arrested** when
  1. disconnected components become less than a threshold, or
  2. a maximum number of iterations is reached
- if  $q_s$  and  $q_g$  can be connected to the **same** component, a solution can be found by **graph search**; else, enhance the PRM by performing more iterations

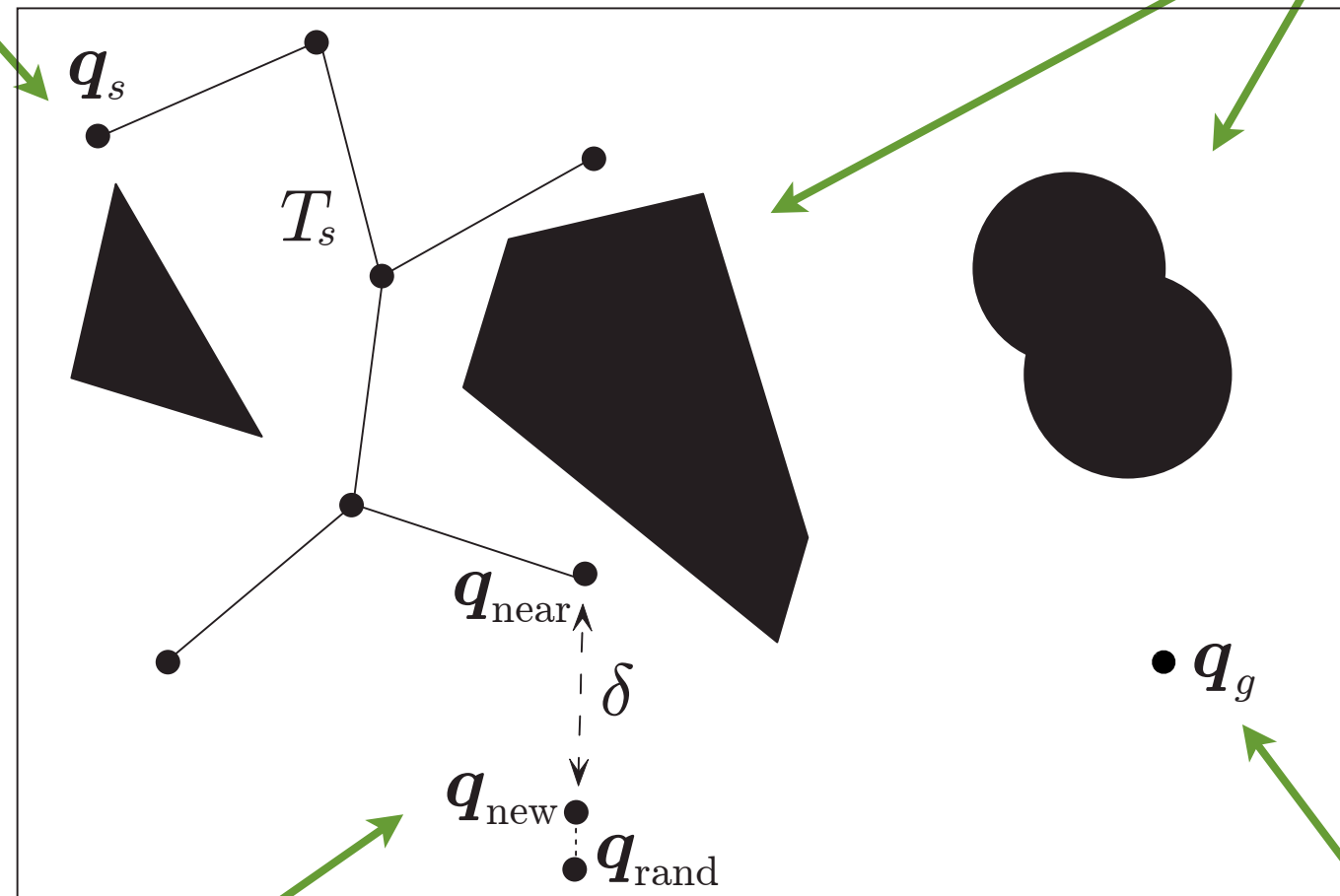
- the PRM method is **probabilistically complete**, i.e., the probability of finding a solution whenever one exists tends to 1 as the execution time tends to  $\infty$ ; and is **multiple-query** (new queries enhance the PRM)
- the main advantage is **speed**; the time PRM needs to find a solution in **high-dimensional spaces** can be orders of magnitude smaller than previous planners
- narrow passages are **critical**; heuristics may be used to design **biased** (non-uniform) probability distributions aimed at increasing sampling in such areas

# RRT (Rapidly-exploring Random Tree)

- basic iteration to build the tree  $T_s$ :
  - root  $T_s$  at  $q_s$
  - generate  $q_{\text{rand}}$  in  $\mathcal{C}$  with **uniform probability distribution**
  - search the tree for the **nearest** configuration  $q_{\text{near}}$
  - choose  $q_{\text{new}}$  at a distance  $\delta$  from  $q_{\text{near}}$  in the direction of  $q_{\text{rand}}$
  - check for **collision**  $q_{\text{new}}$  and the segment from  $q_{\text{near}}$  to  $q_{\text{new}}$
  - if check is negative, add  $q_{\text{new}}$  to  $T_s$  (**expansion**)
- the chosen **metric** in  $\mathcal{C}$  plays a role in identifying  $q_{\text{near}}$
- $T_s$  rapidly covers  $\mathcal{C}_{\text{free}}$  because the expansion is biased towards **unexplored** areas (actually, towards larger Voronoi regions)

tree is  
**rooted** at  $q_s$

$\mathcal{C}$ -obstacles are  
**never** computed

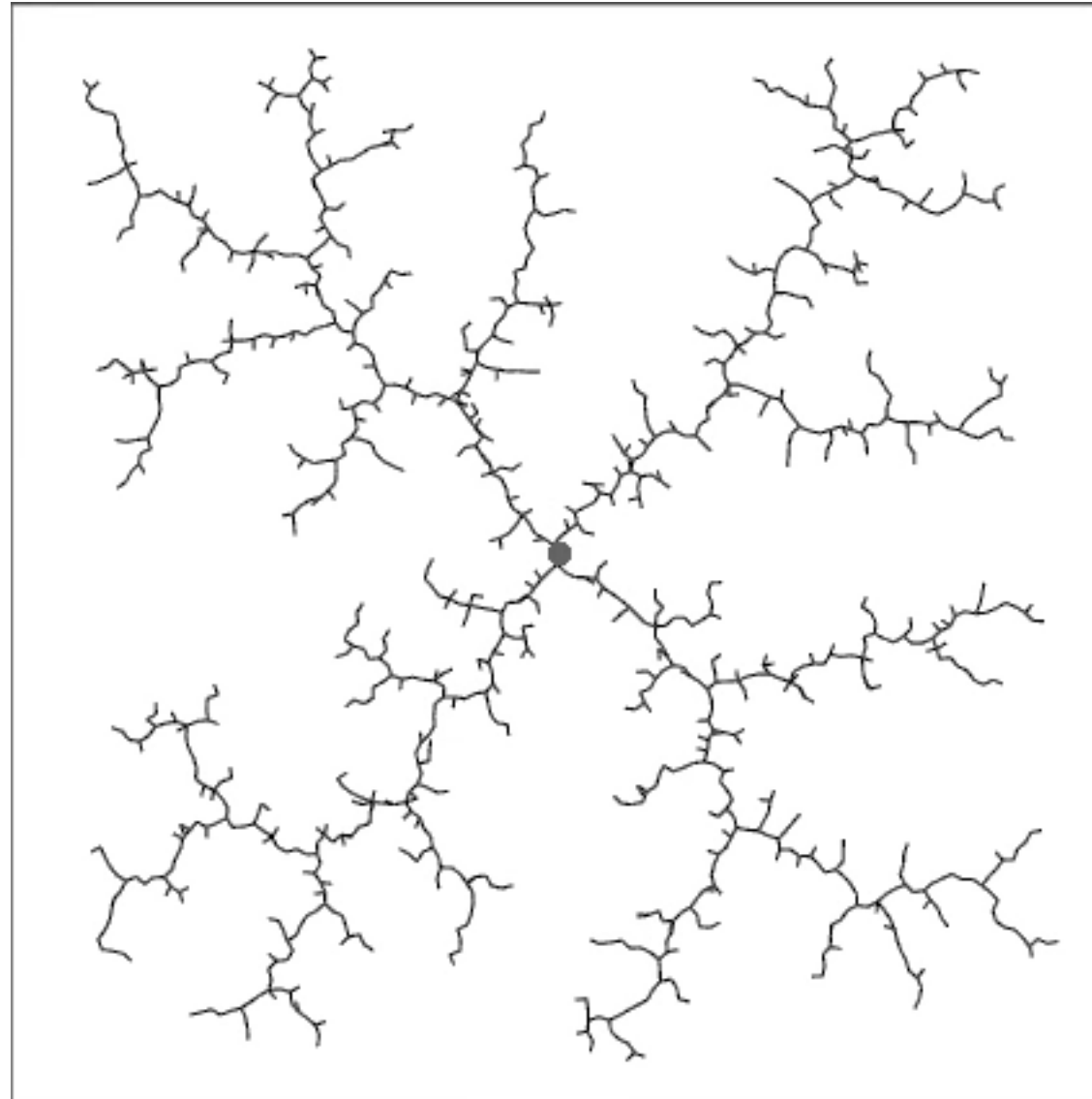


tree  
**expansion**

**no bias**  
towards  $q_g$

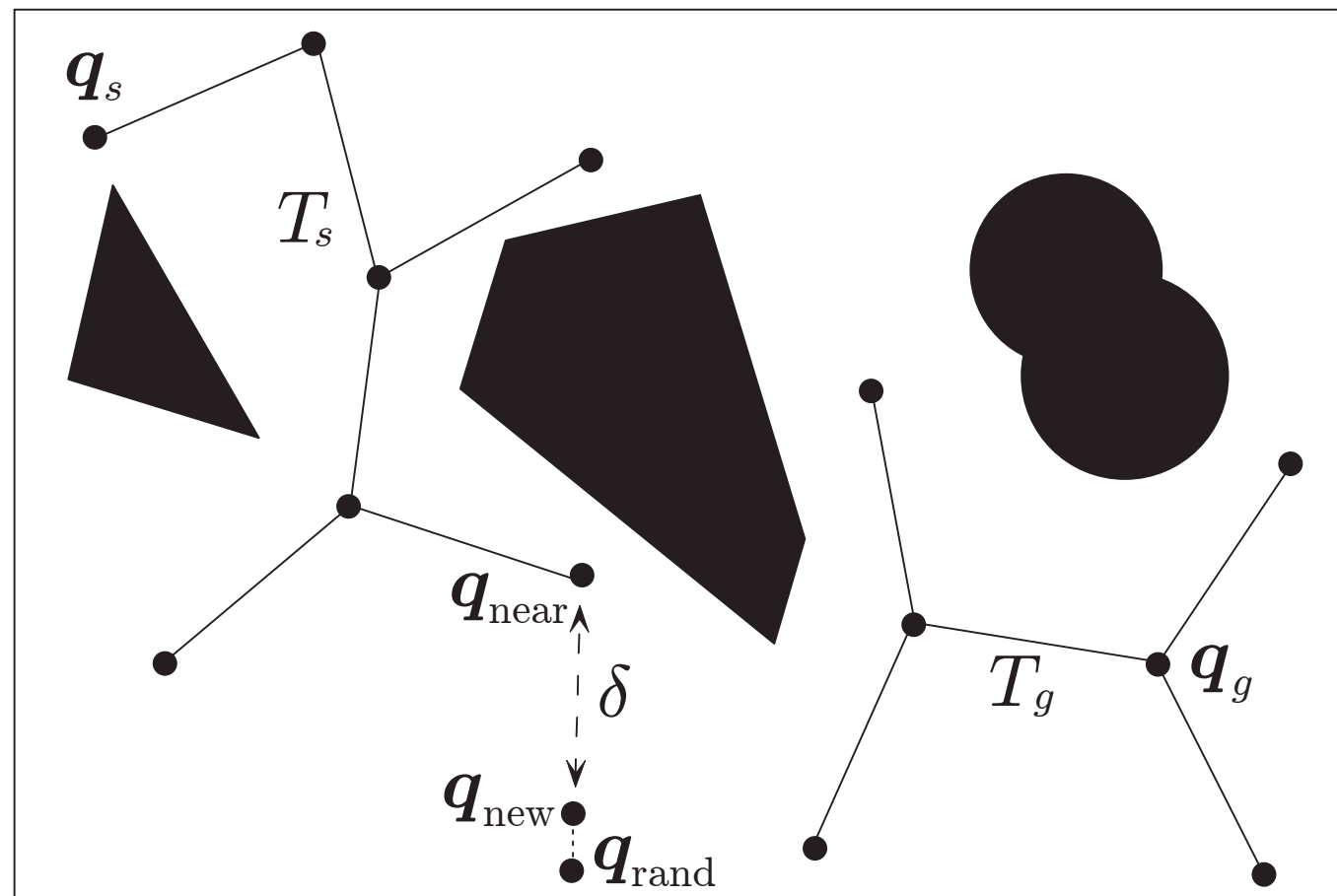


# RRT in empty 2D space



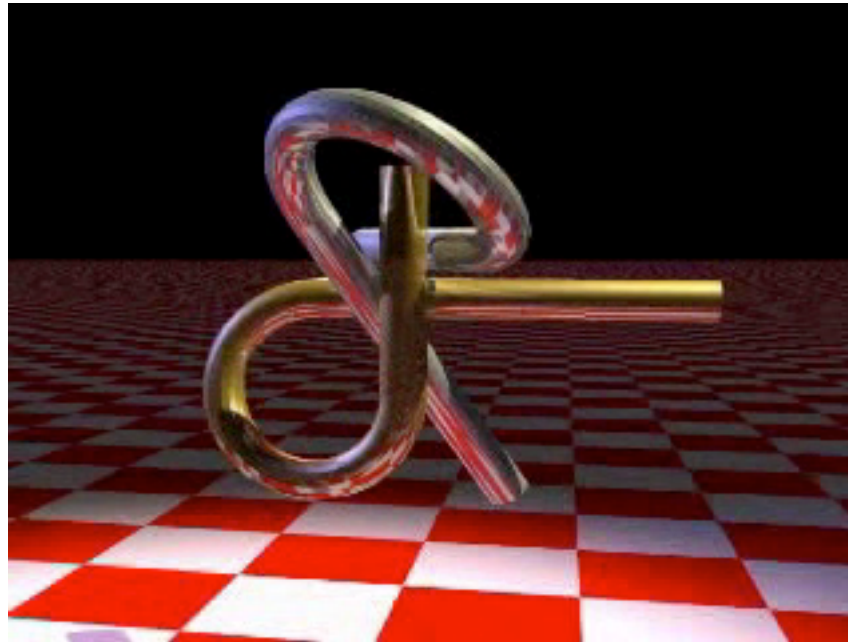
quickly **explores all areas**, much more efficiently than other simple strategies, e.g., random walks

- to introduce a **bias towards  $q_g$** , one may grow **two** trees  $T_s$  and  $T_g$ , respectively rooted at  $q_s$  and  $q_g$  (**bidirectional RRT**)
- alternate expansion and **connection** phases: use the last generated  $q_{\text{new}}$  of  $T_s$  as a  $q_{\text{rand}}$  for  $T_g$ , and then repeat switching the roles of  $T_s$  and  $T_g$



- bidirectional RRT is **probabilistically complete** and **single-query** (trees are rooted at  $q_s$  and  $q_g$ , and in any case new queries may require significant work)
- many variations are possible: e.g., one may use an **adaptive stepsize**  $\delta$  to speed up motion in wide open areas (**greedy** exploration)
- can be modified to address many **extensions** of the canonical planning problem, e.g., moving obstacles, nonholonomic constraints, manipulation planning

# a benchmark problem: the Alpha Puzzle



- 6-dof configuration space + narrow passages
- solved by bidirectional RRT in few mins (average)
- in practice, this problem is not solvable by classical methods such as retraction or cell decomposition

# RRT: extension to nonholonomic robots

- motion planning for a **unicycle** in  $\mathcal{C} = \mathbb{R}^2 \times SO(2)$

$$\begin{pmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{pmatrix} = \begin{pmatrix} \cos \theta \\ \sin \theta \\ 0 \end{pmatrix} v + \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \omega$$

- linear paths in  $\mathcal{C}$  such as those used to connect  $q_{\text{near}}$  to  $q_{\text{rand}}$  are **not admissible** in general
- one possibility is to use **motion primitives**, i.e., a finite set of admissible local paths, produced by a specific choice of the velocity inputs

- for example, one may use (**Dubins car**)

$$v = \bar{v} \quad \omega = \{-\bar{\omega}, 0, \bar{\omega}\} \quad t \in [0, \Delta]$$

resulting in 3 possible paths in forward motion

- the algorithm is the same with the only difference that  $q_{\text{new}}$  is generated from  $q_{\text{near}}$  selecting **one of the possible** paths (either randomly or as the one that leads the unicycle closer to  $q_{\text{rand}}$ )
- if  $q_g$  can be reached from  $q_s$  with a collision-free **concatenation** of primitives, the probability that a solution is found tends to 1 as the time tends to  $\infty$

primitives

solution path made  
by concatenation

