

Sampling-based Motion Planning (Pre Lecture)

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Outline

Sampling-Based Motion Planning

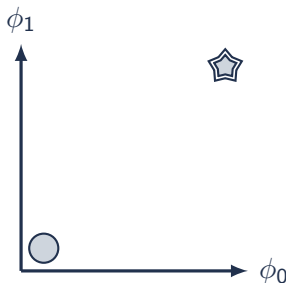
Metric Spaces

Rapidly-Exploring Random Trees (RRT)

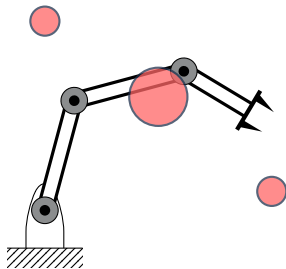
Probabilistic Roadmaps (PRM)

Overview of Sampling-based Motion Planning

1. Start from initial configuration
2. Sample new states in the configuration space
3. Add valid states to tree or graph
4. Terminate when:
 - ▶ We find a path to the goal in the tree/graph
 - ▶ We exhaust a timeout



Configuration Space



- ▶ Configuration Space \mathcal{Q}
- ▶ Obstacle Region: $\mathcal{Q}_{\text{obs}} = \{q \in \mathcal{Q} \mid \neg \text{is-valid}(q)\}$
- ▶ Free Space: $\mathcal{Q}_{\text{free}} = \{q \in \mathcal{Q} \mid \text{is-valid}(q)\}$
- ▶ Total Space:
 - ▶ $\mathcal{Q}_{\text{obs}} \cap \mathcal{Q}_{\text{free}} = \emptyset$
 - ▶ $\mathcal{Q}_{\text{obs}} \cup \mathcal{Q}_{\text{free}} = \mathcal{Q}$

Generally: no explicit representation of \mathcal{Q}_{obs} , $\mathcal{Q}_{\text{free}}$

Fundamental Abstractions

Configuration Space: Set of possible configurations

► \mathcal{Q}

Configuration Sampler: Generates candidate configurations

► $() \mapsto \mathcal{Q}$

Validity Checker: Is some configuration in $\mathcal{Q}_{\text{free}}$?

► $\underbrace{\mathcal{Q}}_{\text{sample}} \mapsto \underbrace{\{0, 1\}}_{\text{valid?}}$

Nearest Neighbors: Closest point in tree/graph to new sample:

► $\underbrace{\mathcal{P}(\mathcal{Q})}_{\text{tree}} \times \underbrace{\mathcal{Q}}_{\text{new sample}} \mapsto \underbrace{\mathcal{Q}}_{\text{nearest neighbor}}$

Sampling-based motion planning generalizes beyond robotics.

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Definition: Metric Space

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A **metric space** is a space \mathcal{Q} equipped with a function $\rho : \mathcal{Q} \times \mathcal{Q} \mapsto \mathbb{R}$ that has the following properties for any $a, b, c \in \mathcal{Q}$

Nonnegative: $\rho(a, b) \geq 0$

(Distances are always greater than zero)

Reflexive: $(\rho(a, b) = 0) \iff (a = b)$

(Distance is zero only for identical elements)

Symmetric: $\rho(a, b) = \rho(b, a)$

(Distance from a to b is the same as from b to a)

Triangle: $\rho(a, b) + \rho(b, c) \geq \rho(a, c)$

(a to b to c cannot be shorter than directly from a to c)

Example: Robot Joint Space and L_p metrics

$$\rho(x, x') = \left(\sum_{i=0}^{n-1} |x_i - x'_i|^p \right)^{\frac{1}{p}}$$

► **Space:**

$\mathcal{Q} \subseteq \mathbb{R}^n$, the set of robot joint positions

► **Euclidean Distance:**

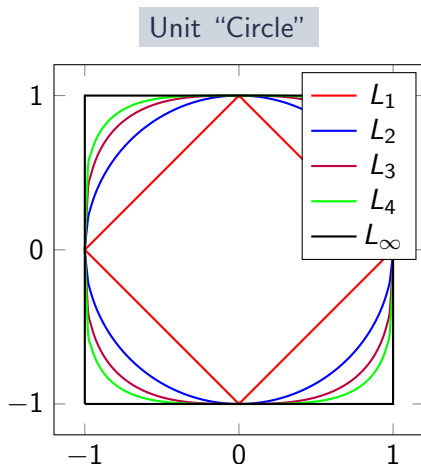
$$L_2(x, x') = \sqrt{\sum_{i=0}^{n-1} (x_i - x'_i)^2}$$

► **Manhattan Distance:**

$$L_1(x, x') = \sum_{i=0}^{n-1} |x_i - x'_i|$$

► **L_∞ :**

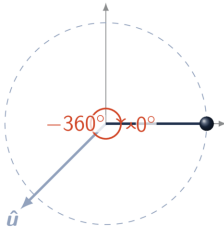
$$L_\infty(x, x') = \max_{0 \leq i \leq n-1} (|x_i - x'_i|)$$



Example: Rotation Metric

Quaternion Double-Cover

Axis-Angle



$$\begin{aligned}\theta &= 0.000 \\ \theta - 2\pi &= -6.280\end{aligned}$$

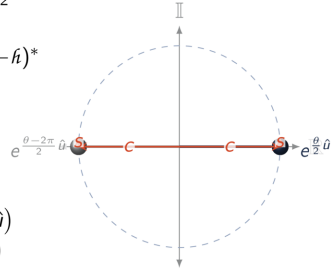
$$\exp\left(\frac{\theta}{2}\hat{u}\right) = \sin\frac{\theta}{2}\hat{u} + \cos\frac{\theta}{2}$$

$$\hat{h} \otimes \vec{p} \otimes \hat{h}^* = (-\hat{h}) \otimes \vec{p} \otimes (-\hat{h})^*$$

$$\begin{aligned}\sin\frac{\theta}{2} &= -\sin\frac{\theta-2\pi}{2} \\ \cos\frac{\theta}{2} &= -\cos\frac{\theta-2\pi}{2}\end{aligned}$$

$$\begin{aligned}\exp\left(\frac{\theta}{2}\hat{u}\right) &= -\exp\left(\frac{\theta-2\pi}{2}\hat{u}\right) \\ \exp\left(\frac{\theta}{2}\hat{u}\right) &\sim \exp\left(\frac{\theta-2\pi}{2}\hat{u}\right)\end{aligned}$$

Quaternion



$$\begin{aligned}0.000 \hat{u} + 1.000 \\ \sim 0.000 \hat{u} + -1.000\end{aligned}$$

Example: Rotation Metric

Distance Functions

$$\exp\left(\frac{\theta}{2}\hat{u}\right) = \sin\frac{\theta}{2}\hat{u} + \cos\frac{\theta}{2} \quad \text{and} \quad \exp\left(\frac{\theta}{2}\hat{u}\right) \sim \exp\left(\frac{\theta - 2\pi}{2}\hat{u}\right)$$

Quaternion Space

- ▶ L_p Norm:
 - ▶ $\min\{\|p - q\|, \|p + q\|\}$
 - ▶ Length of 4D line segment
- ▶ 4D angle:
 - ▶ $\min\{\cos^{-1}(q \bullet p), \cos^{-1}(-q \bullet p)\}$
 - ▶ Angle between 4D quaternions

Log Space

$$\min\{\|\ln(q \otimes p^*)\|, \|\ln(-q \otimes p^*)\|\}$$

- ▶ $p \otimes h_r = q \rightsquigarrow h_r = q \otimes p^*$
- ▶ Matrix: $\|\ln(\mathbf{PQ}^{-1})\|$

Exercise: Transformation Metric

Dual Quaternion: $\rho(h_1 + d_1\epsilon, h_2 + d_2\epsilon)$

Transformation Matrix: $\rho\left(\begin{bmatrix} \mathbf{R}_1 & \mathbf{v}_1 \\ \mathbf{0} & \mathbf{1} \end{bmatrix}, \begin{bmatrix} \mathbf{R}_2 & \mathbf{v}_2 \\ \mathbf{0} & \mathbf{1} \end{bmatrix}\right)$

Outline

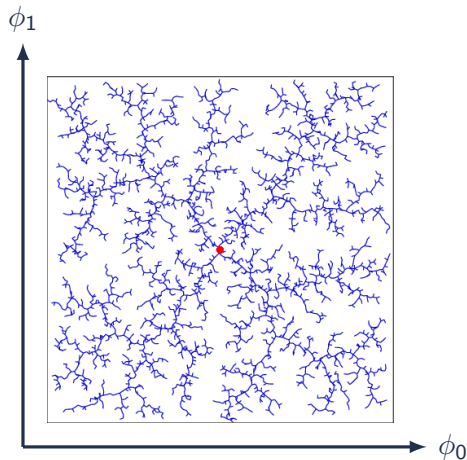
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RRT Illustration



<http://msl.cs.uiuc.edu/rrt/>

RRT Algorithm

Algorithm 1: Rapidly-Exploring Random Tree

Input: $\phi_0 \in \mathcal{Q}$: start

Input: $\Phi_g \subseteq \mathcal{Q}$: goal set

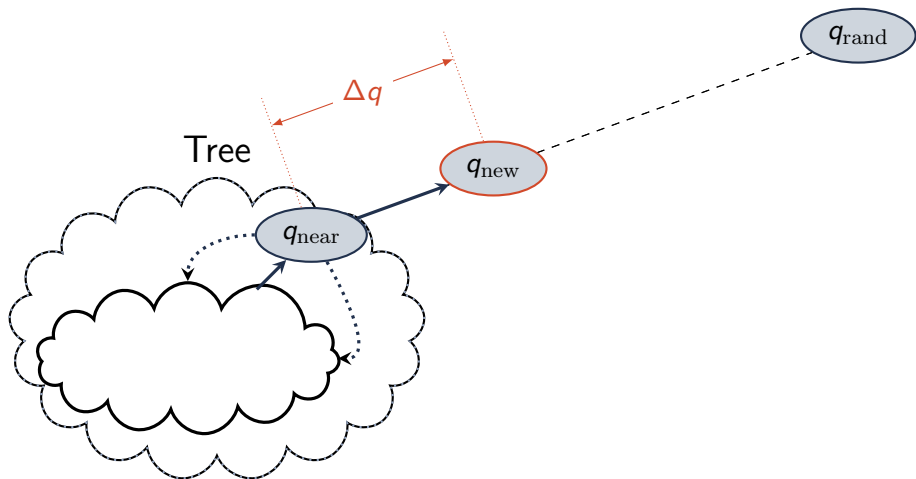
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1  $V \leftarrow \{\phi_0\};$ 
2  $E \leftarrow \emptyset;$ 
3 for  $k \leftarrow 0$  to LIMIT do
4    $q_{\text{rand}} \leftarrow \text{sample}();$ 
5    $q_{\text{near}} \leftarrow \text{nearest-neighbor}(V, q_{\text{rand}});$ 
6    $q_{\text{new}} \leftarrow \text{new-conf}(q_{\text{near}}, q_{\text{new}});$ 
7   if  $\text{valid}(q_{\text{new}})$  then
8     if  $\exists q \in \Phi_g, \text{dist}(q, q_{\text{new}}) < \epsilon$  then
9       return path from  $q_0$  to  $q_{\text{new}};$ 
10     $V \leftarrow V \cup \{q_{\text{new}}\};$ 
11     $E \leftarrow E \cup \{q_{\text{near}} \rightarrow q_{\text{new}}\};$ 
12 return TIMEOUT;
```

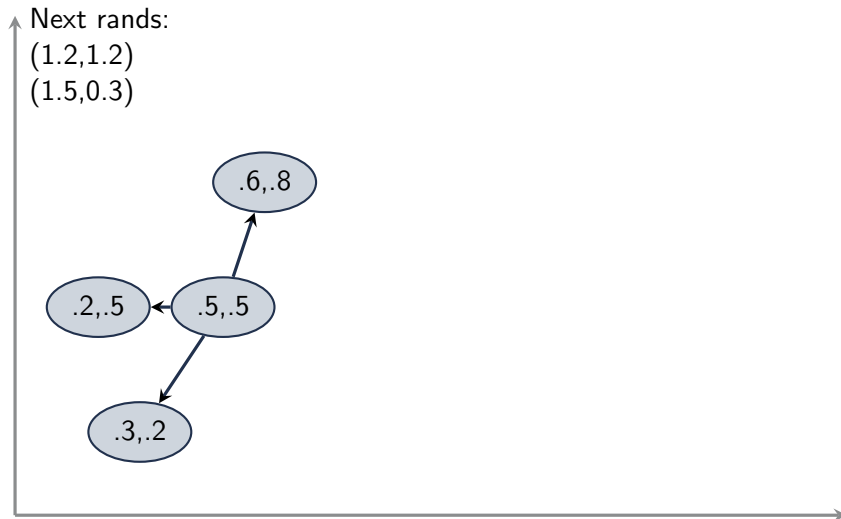
Subroutines

- ▶ sample
- ▶ nearest-neighbor
- ▶ new-conf
- ▶ valid
- ▶ dist

RRT Step Illustration



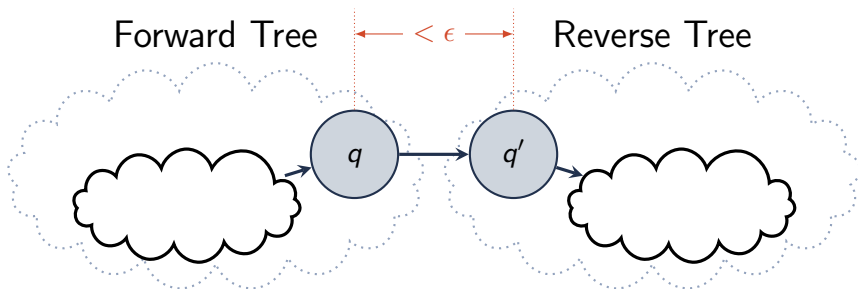
Exercise: RRT-Step



RRT-Connect

Bi-directional RRT

1. Construct one RRT, rooted at the start
2. Construct second RRT, rooted at the goal
3. Terminate when the two trees connect



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PRM Overview

Preprocessing: Construct the Roadmap

1. Sample new configurations
2. Connect to “neighboring” configurations in roadmap

Query: Search the Roadmap

1. Find path in roadmap via discrete search (e.g., A^*)
2. Connect subsequent configurations with “local planner” (interpolate)

PRM Construction

Algorithm

Algorithm 2: Construct Probabilistic Roadmap

```

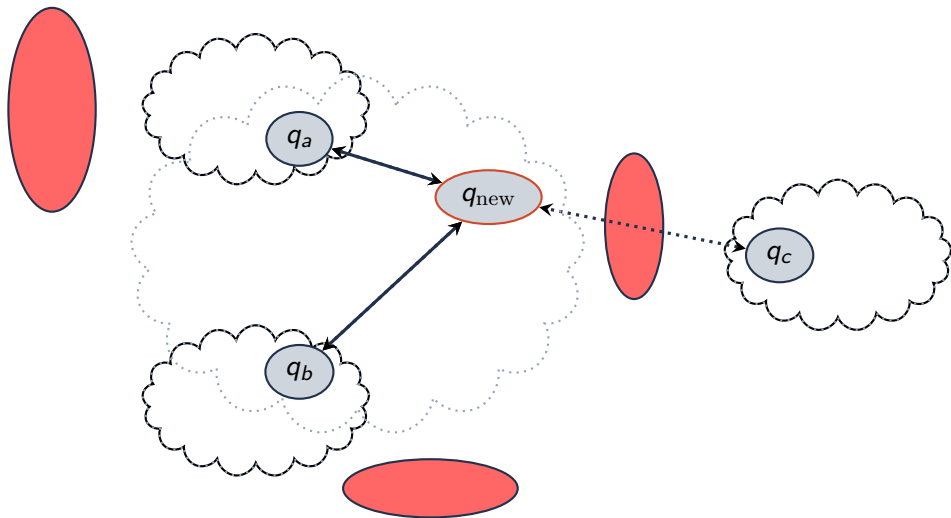
1   $V \leftarrow \emptyset$ ;
2   $E \leftarrow \emptyset$ ;
3   $k \leftarrow 0$ ;
4  while  $k < \text{LIMIT}$  do
5       $q_{\text{new}} \leftarrow \text{sample}()$ ;
6      if  $\text{valid}(q_{\text{new}})$  then
7           $V \leftarrow V \cup \{q_{\text{new}}\}$ ;
8           $k \leftarrow k + 1$ ;
9          foreach  $q \in \text{neighborhood}(q_{\text{new}})$  do
10             if  $\neg \text{connected}(q, q_{\text{new}}) \wedge$ 
11                  $\text{connect}(q, q_{\text{new}})$  then
12                  $E \leftarrow E \cup \{q_{\text{new}} \rightarrow q\}$ ;
12 return  $(V, E)$ ;

```

- ▶ Neighbors:
 - ▶ K-nearest neighbors (KNN)
 - ▶ KNN from connected-components of V
 - ▶ Radius (within ϵ distance)
- ▶ Connected Components:
 - ▶ Optional: do we want different paths?
 - ▶ Efficiency: Track as edges are added
- ▶ Connect: valid path exists
 - ▶ E.g., linearly interpolate and check validity

PRM Construction

Illustration



Practical Considerations

- ▶ Nearest Neighbors: Kd-tree (partitioning) or brute force (for high-dof)
- ▶ Collision Checking: Bounding-boxes then decompose into triangles
- ▶ Paths: Smooth and Shortcut
- ▶ Hundreds of variations on RRT, PRM, and similar methods
- ▶ RRT-Connect usually works pretty well