

RMP-Sample Based Motion Planning

☰ Key words	<div>connectivity</div> <div>expansiveness</div> <div>multiple query</div> <div>probabilistic roadmap</div> <div>single query</div> <div>smoothing</div>
☰ Status	<div>note complete</div>

Dis

Combinatorial approaches:

- Advantages: **complete algorithm**
- Disadvantages:
 - **computational expensive**, DOF \uparrow , runtime increases exponentially \uparrow
 - construction of **whole configuration space**(collision tests) \rightarrow for larger DOF very expensive
 - uses heuristics, not reliable. (potential field)

\rightarrow Solution: **Sample based methods!!!**

Sample based Methods

- **probabilistic completeness**: a solution path can be **found with high probability**.
- ▼ Distribution of sample points: **uniform random distribution**
 - on which parameters?

▼ **Parametrization** of environment to get a **connected roadmap**

Probabilistic Roadmap (PRM) — Multiple-query sampling for higher dimensions solving multiple-query problems

- **multiple query sampling**

▼ **assumptions:**

- **static** environment, arbitrary obstacles
- **many queries** processed in same environment
- **unknown start & goal**, can be anywhere in free space
- repeated tasks

—> **sample over the whole free space** necessary!!

▼ **applications:**

- navigation in static environment
 - robot manipulator arm in a workcell (parts from different start, need to be moved to different goals)
 - playing chess
- **Process:**

▼ **Offline: Construct PRM**

1. **sampling (nodes): uniformly sample n milestones** in the config-space & **collision test**
 - a. in free space: kept
 - b. collision with obstacle: try again
2. **connection (edges): connect milestones through linear local path. Collision test** along the linear local path by **dense sampling**
 - a. if no collision from each point: connect 2 nodes
 - b. collision:

▼ Online: Query

1. given start/goal point, **connect** them to the **nearest milestone**(node) in PRM by **random walk**.
 2. **graph search**: find a path based on cost function (shortest path, least node, etc)
 3. termination:
 - a. if path returned: always correct
 - b. no path returned: error, but hope correct with high probability
- Problem: small connected components, **low connectivity of the whole graph**

▼ Solutions to increase connectivity:

- increase #samples

▼ random walk

- collision during linear connection: walk in **random direction**
- when hits an obstacle, random walk again **up to a maximum distance** d_{max} → create **new milestone**.
- try to connect. repeat until 2 subgraphs connect.

→ edges can be no longer straight line, still linear

→ might get you **through narrow passages /opening**

→ resulting path could be very zig-zag: **smoothing necessary!!!**

▼ path correction

- collision when connecting 2 nodes: push the line towards free space → **find boundary**
- sample points along the line: for each point, extend **in random direction**
- **binary search**: collision test in by halving the distance **until boundary** —> **create new milestone**.
- connect the milestones with subgraphs.

▼ sample around obstacles: OBPRM

▼ Smoothing of zig-zag paths

▼ motivation:

- new nodes that increase connectivity create **zig-zag path**, chaotic & complicated
- frequent **change of configuration and velocity (negative to positive)**

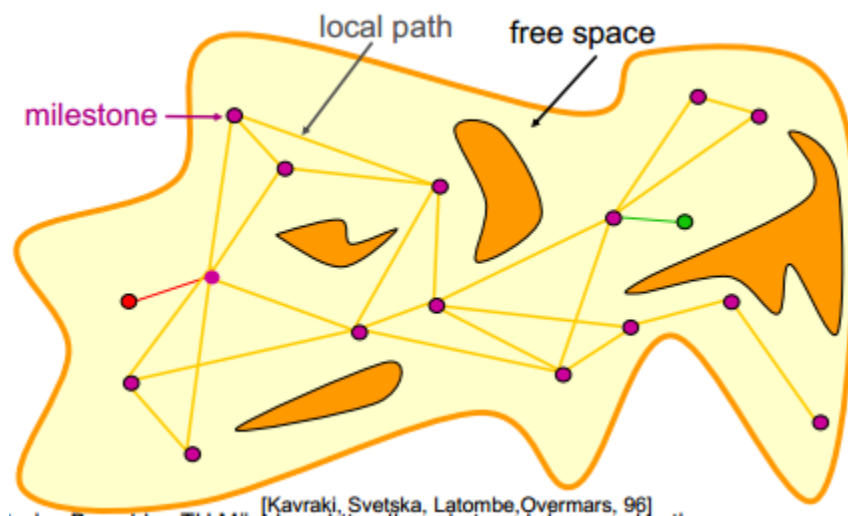
—> **smoothing necessary!!!!**

▼ Process

1. connect start to goal — **supporting line**
2. if collision, find the **most distant node** to the line.
3. from that node, connects to goal — **supporting line**
4. repeat, find the **most distant node**, until it reaches the goal.

—> Path: sequence of these most distant nodes.

- Preservation of **planning topology: homotopic path**
 - after path correction(smoothing), **homotopy** ensures the **space topology** gets visited from the **same side as planned**.
 - it **doesn't intersect any obstacles**. no jumping through an obstacle.



Probabilistic Roadmap (PRM) — Multiple-query sampling solving single-query problems

- **single query problem:**
 - **known start & goal**, not far away from each other.
 - **one-time** task
- **multiple-query sampling solution:**
 - add start & goal with all other uniformly sampled distributed nodes
 - build edges

—> sampling the whole free space uniformly doesn't make sense! —> resource waste

—> **single-query sampling!!**

Single-query sampling solving single-query problems

▼ Process:

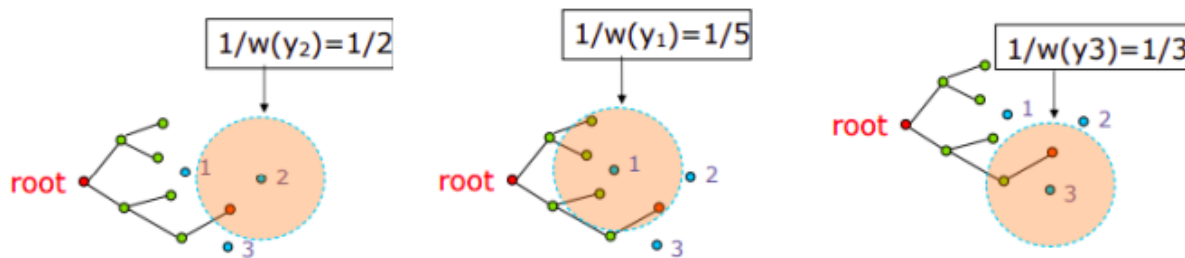
1. **expansion:**
 - a. simultaneously **grow 2 trees from start & goal** configuration
 - b. **randomly sample n nodes** around existing nodes —> **weighted sampling**
2. **connection:** check if **distance** of child nodes between 2 trees is **below threshold** —> **linear dense sampling** —> **connect** if no collision
3. iterates between expansion and connection until
 - a. 2 trees connected
 - b. max. number of expansion/connection steps reached



Expansion + Connection

- Problem at sampling:
 - child nodes can **expand back** to the parent nodes (back to known areas)
 - if **obstacle**, the tree will **bounce back** to parent nodes.
 - > unable to go through narrow passages.
- ▼ Solution at **Expansion: weighted sampling of nodes**
 - Weight: $w(x)$ — **number of existing nodes** (including itself) within **radius d**
 - 1. pick a node x with probability $\frac{1}{w(x)}$
 - 2. randomly sample k points around x within radius d
 - 3. for each sample y , calculate $w(y)$ and **sampling probability** $\frac{1}{w(y)}$
 - 4. add y into the tree, if
 - a. y has higher probability
 - b. collision free from y to x (linear collision test)
 - c. y can see x

—> nodes **well distributed** over unknown spaces, also able to **expand to obstacle** or through **narrow passages!!!**



▼ applications:

- remove parts from car: a collision free path
- piano mover
- start & goal close to each other, one-time tasks

Evaluation of PRM: $(\epsilon, \alpha, \beta)$ - Expansiveness

- **coverage**: milestones are distributed such that **almost any point** of config-space can be **connected to one milestone** with a **straight line**. (doesn't have to be connected)
- **connectivity**: a **one-to-one correspondence** between the **connected components** of the roadmap and those of F

—> difficult in narrow passages

▼ **expansiveness**: characterize both **coverage** and **connectivity**

- **visibility set** of q : all configurations seen by q by straight line.
 - **minimum visibility ϵ** : **each configuration** sees **at least ϵ fraction** of the **free space**
- > find the configuration q with minimum visibility

$$\epsilon = \frac{S}{F_{\text{freespace}}} \in (0, 1]$$

- **β -lookout set of S** — B, subset of the confined space S: the **area from the confined space S**(minimum visibility) where it **can see the rest of the free space**
- **lookout of B** — β : **each configuration** in B can see **at least β fraction of freespace $F \setminus S$**

$$\beta = \frac{C}{F \setminus S} \in (0, 1]$$

- α : the fraction of B to the confined space S

$$\alpha = \frac{B}{S}$$

- free space F is $(\epsilon, \alpha, \beta)$ - **expansive**:

- F is **ϵ -good**,
- for each subset S in F, its **β -lookout is at least α fraction of S**.

—> $\epsilon, \alpha, \beta \uparrow$, construction cost \downarrow for good connectivity and coverage.

—> **high probability** in getting a **linking sequence** that covers a **large fraction of free space**

Theorem: a roadmap of $n = \frac{8 \ln(8/\epsilon \cdot \alpha \cdot \gamma)}{\epsilon \cdot \alpha} + \frac{3}{\beta}$ milestones has the correct connectivity with **at least $1 - \gamma$ success rate**.

—> exponential dependency of success and n: **increase n linearly** will result in **exponential improvement of**

- **success**
- **connectivity**
- **coverage**

- basic PRM: uniform random distribution
- not fast for single query
- single query sampling:
 - EST
 - RRT