

Last lecture

- Multiple-query PRM
- Lazy PRM (single-query PRM)

Single-Query PRM

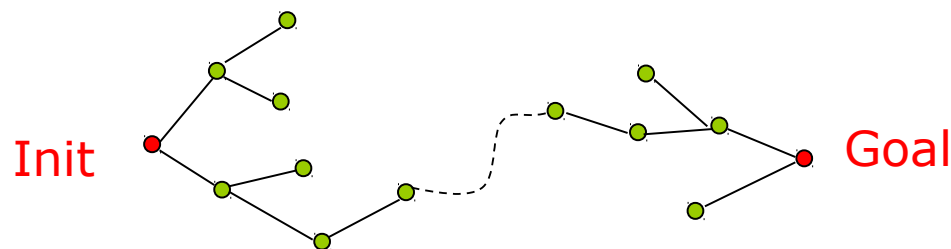


Randomized expansion

- *Path Planning in Expansive Configuration Spaces*,
D. Hsu, J.C. Latombe, & R. Motwani, 1999.

Overview

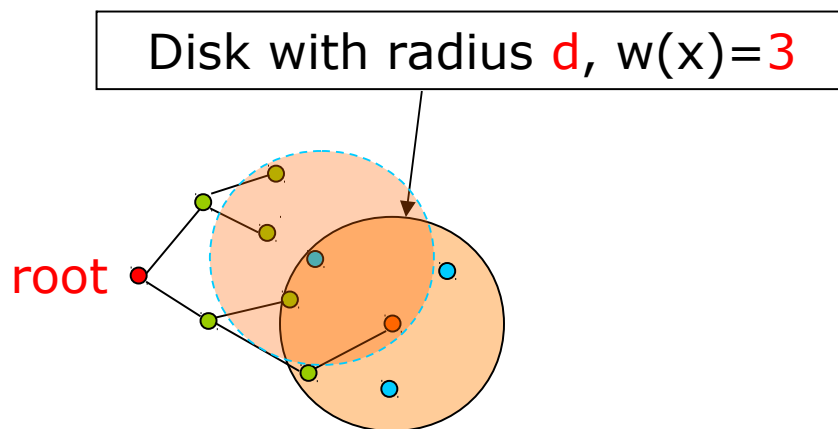
1. Grow two trees from Init position and Goal configurations.
2. Randomly sample nodes around existing nodes.
3. Connect a node in the tree rooted at Init to a node in the tree rooted at the Goal.



Expansion + Connection

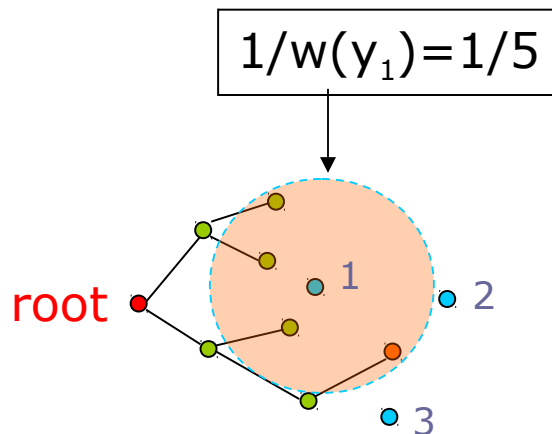
Expansion

1. Pick a node x with probability $1/w(x)$.
2. Randomly sample k points around x .
3. For each sample y , calculate $w(y)$, which gives probability $1/w(y)$.



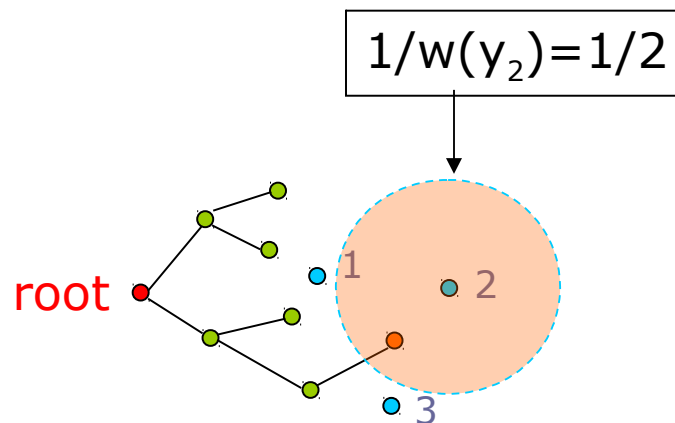
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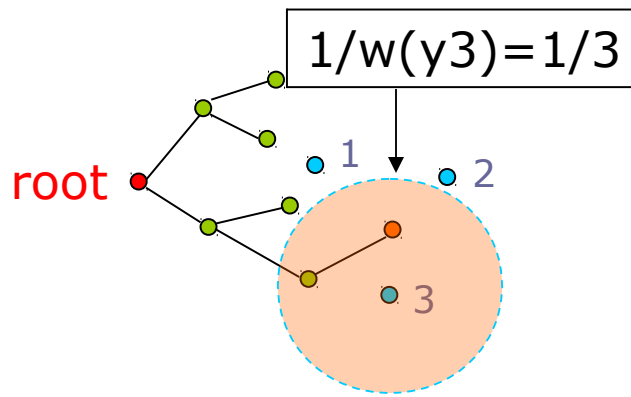
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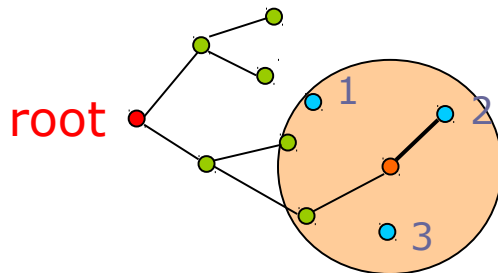
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Expansion

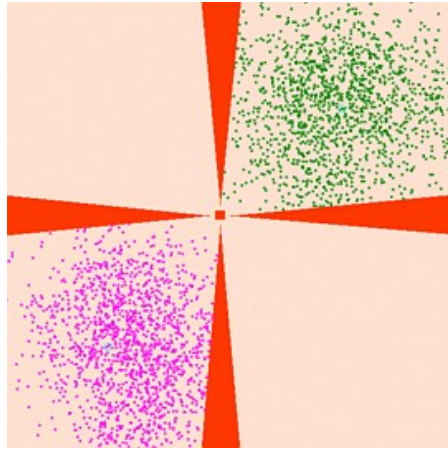
1. Pick a node x with probability $1/w(x)$.
2. Randomly sample k points around x .
3. For each sample y , calculate $w(y)$, which gives probability $1/w(y)$. If y
(a) has higher probability; (b) collision free; (c) can see x
then add y into the tree.



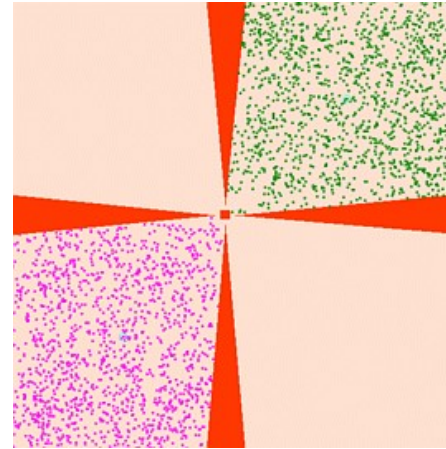
Sampling distribution

- Weight $w(x)$ = no. of neighbors
- Roughly $\Pr(x) \sim 1 / w(x)$

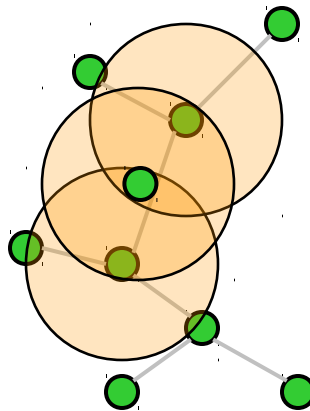
Effect of weighting



unweighted sampling

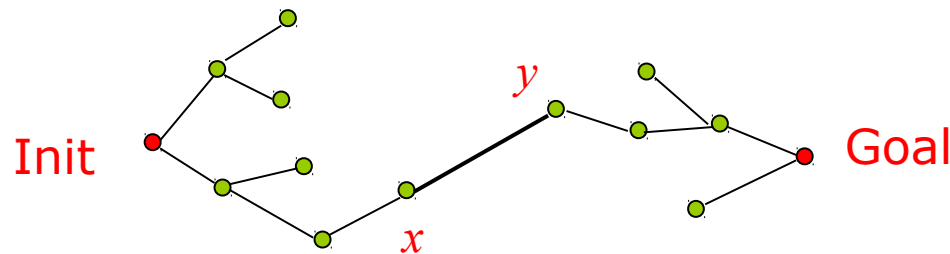


weighted sampling



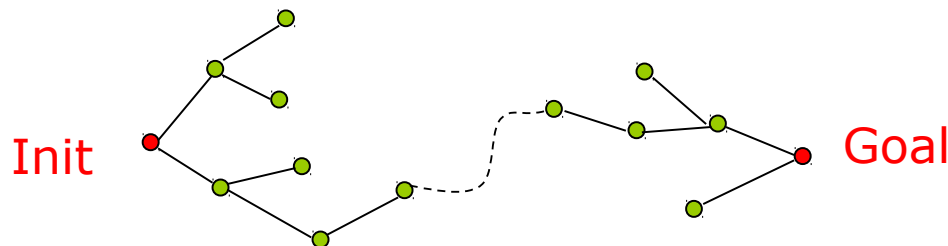
Connection

- If a pair of nodes (*i.e.*, x in Init tree and y in Goal tree) and $\text{distance}(x,y) < L$, check if x can see y
YES, then connect x and y

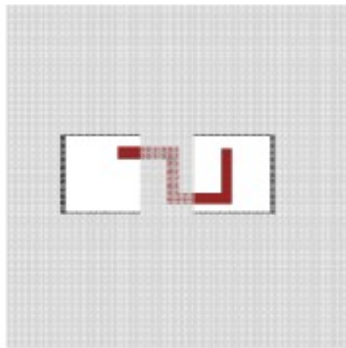


Termination condition

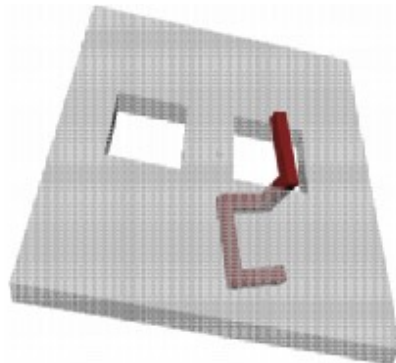
- The program iterates between **Expansion** and **Connection**, until
 - two trees are connected, or
 - max number of expansion & connection steps is reached



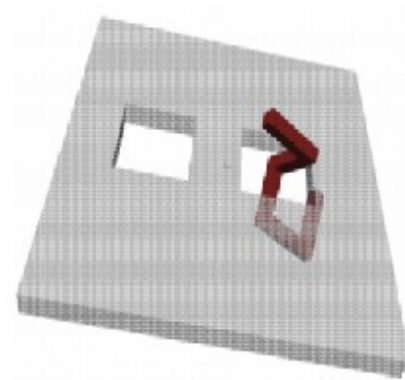
Computed example



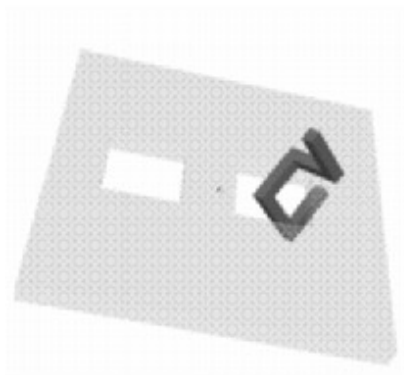
(a)



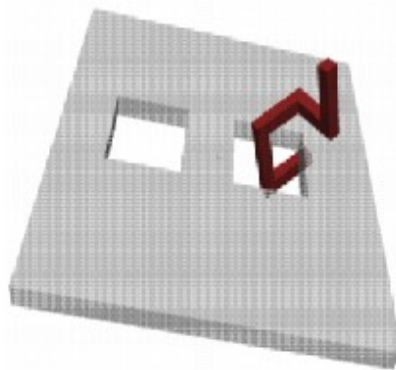
(b)



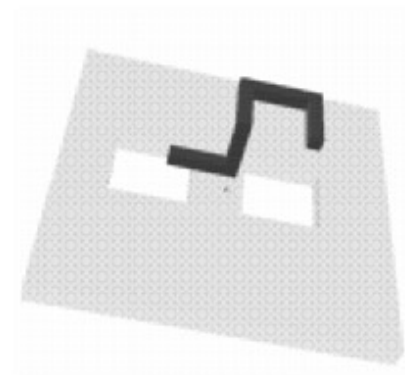
(c)



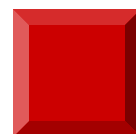
(d)



(f)



(g)



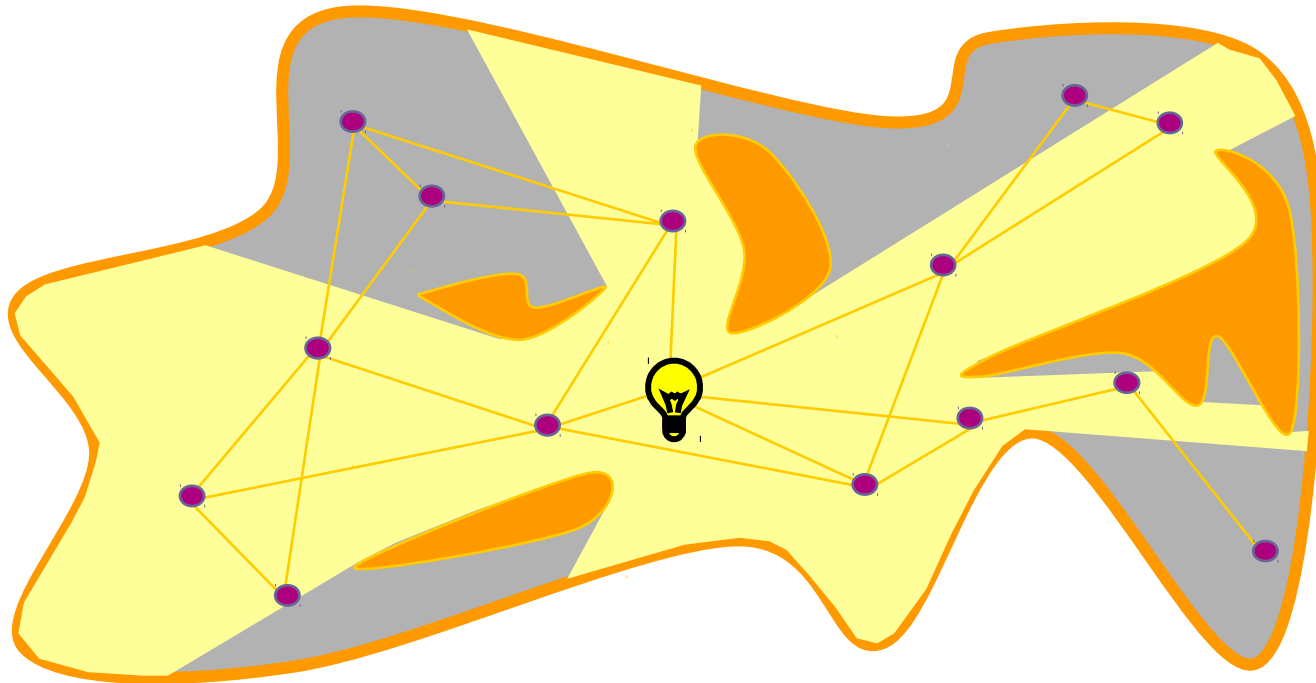
Expansive Spaces



Analysis of Probabilistic Roadmaps

Issues of probabilistic roadmaps

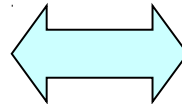
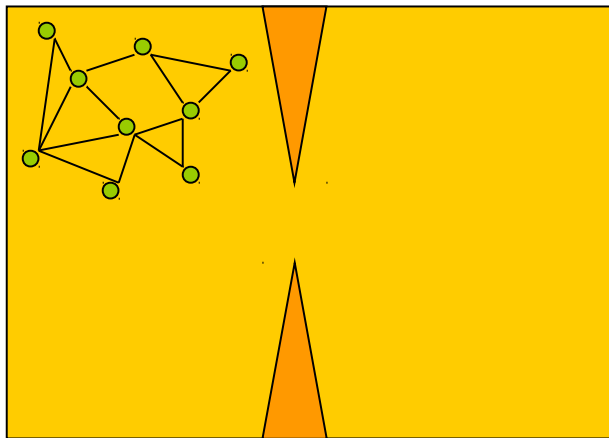
- Coverage
- Connectivity



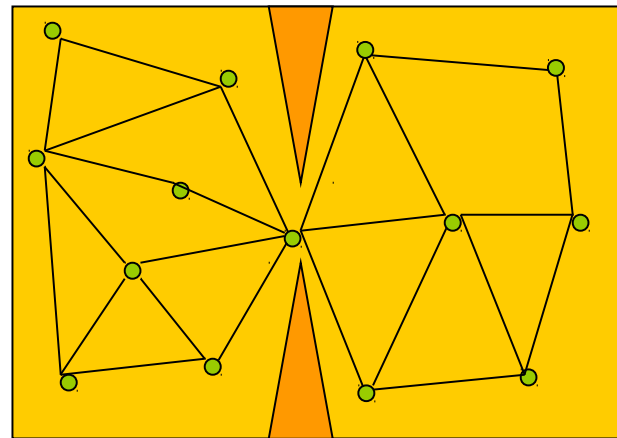
Is the coverage adequate?

- It means that milestones are distributed such that almost any point of the configuration space can be connected by a straight line segment to one milestone.

Bad

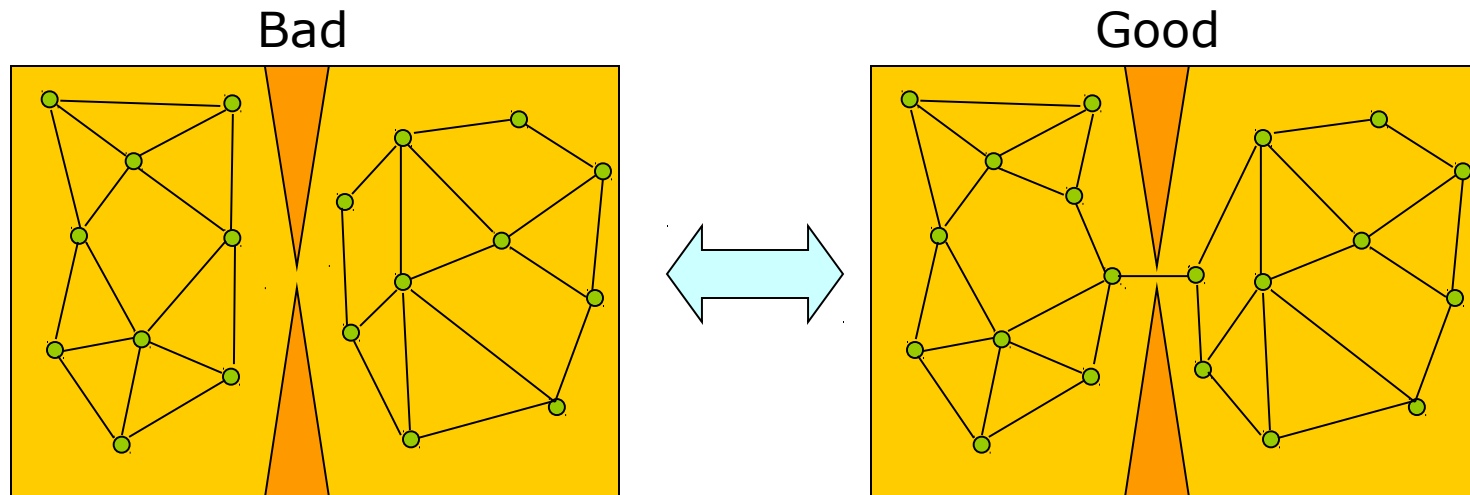


Good



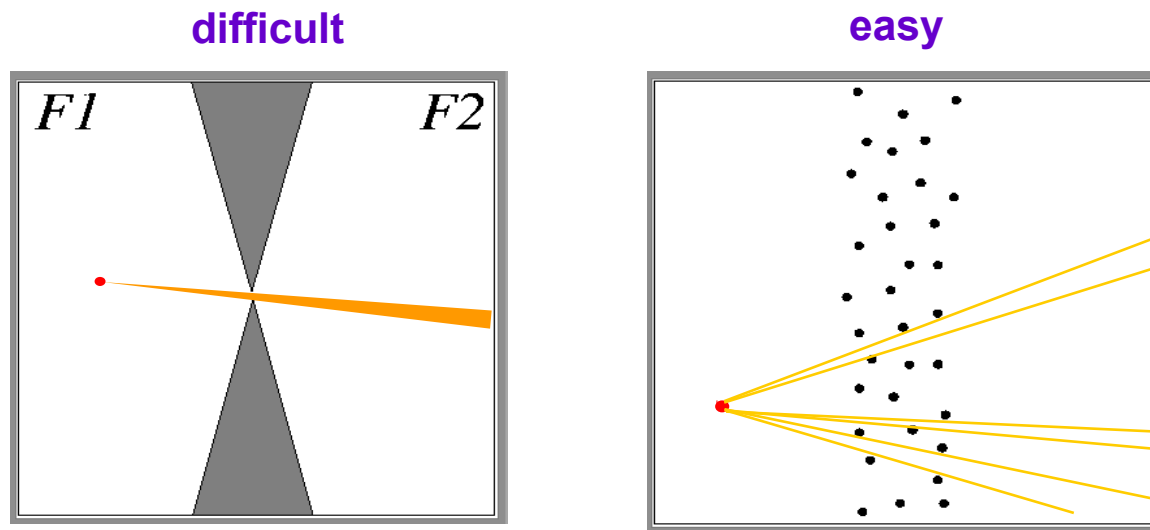
Connectivity

- There should be a one-to-one correspondence between the connected components of the roadmap and those of F .



Narrow passages

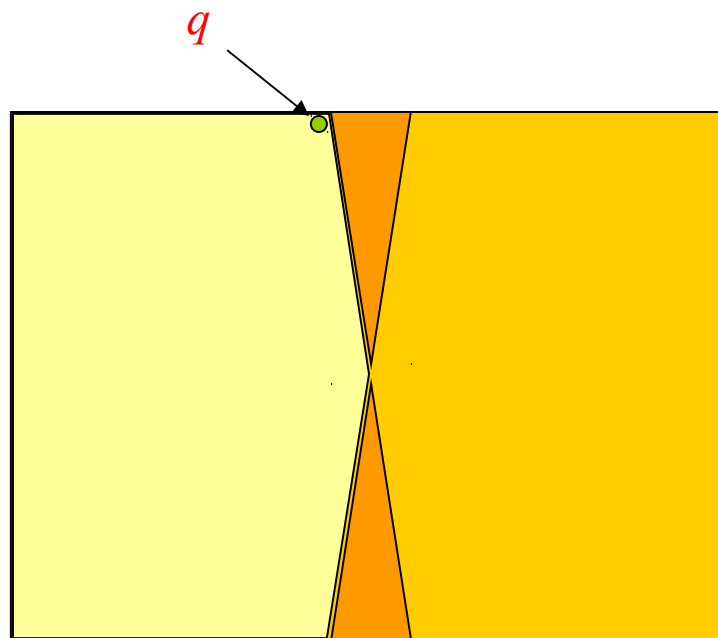
- Connectivity is difficult to capture when there are narrow passages.
- Narrow passages are difficult to define.



Characterize coverage & connectivity? → **Expansiveness**

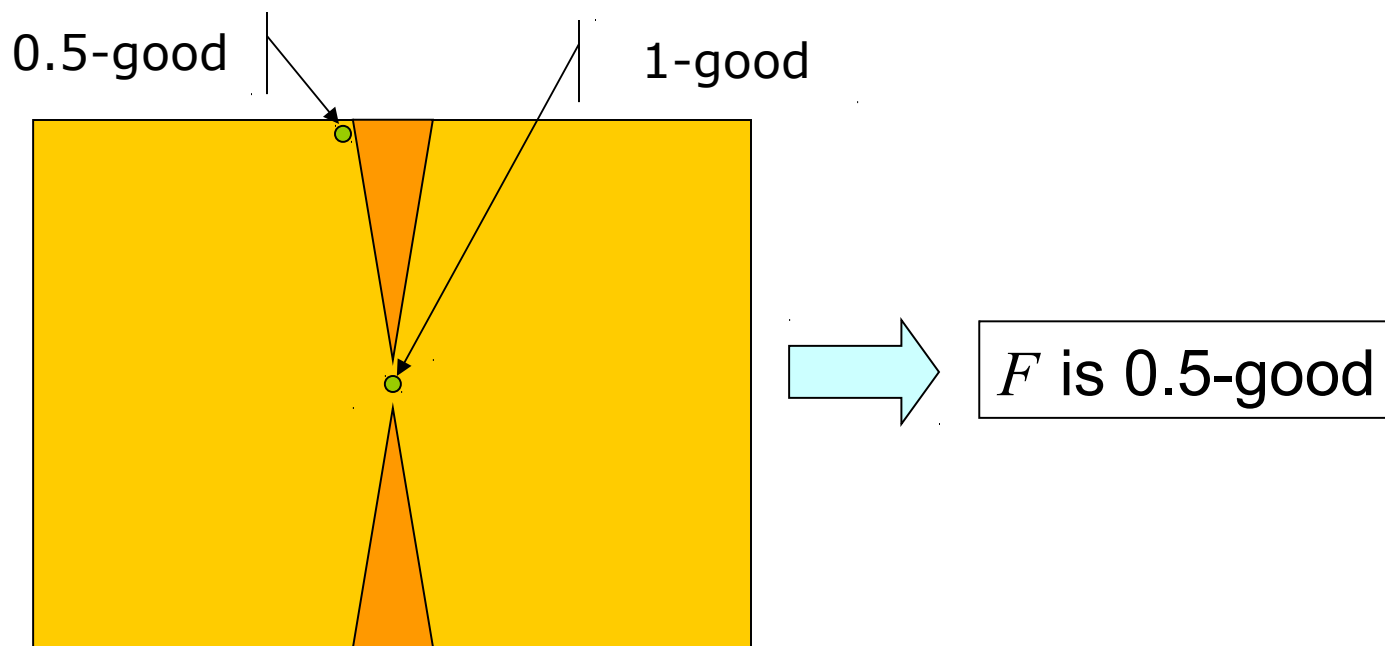
Definition: visibility set

- Visibility set of q
 - All configurations in F that can be connected to q by a straight-line path in F
 - All configurations seen by q



Definition: ϵ -good

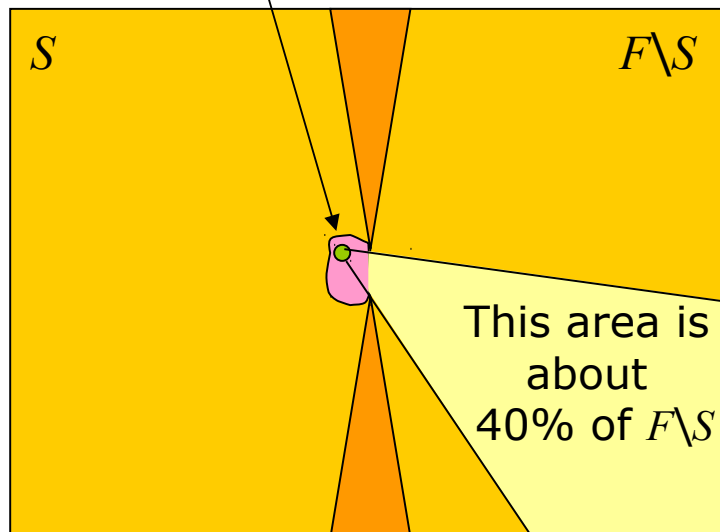
- Every free configuration sees at least ϵ fraction of the free space, ϵ in $(0,1]$.



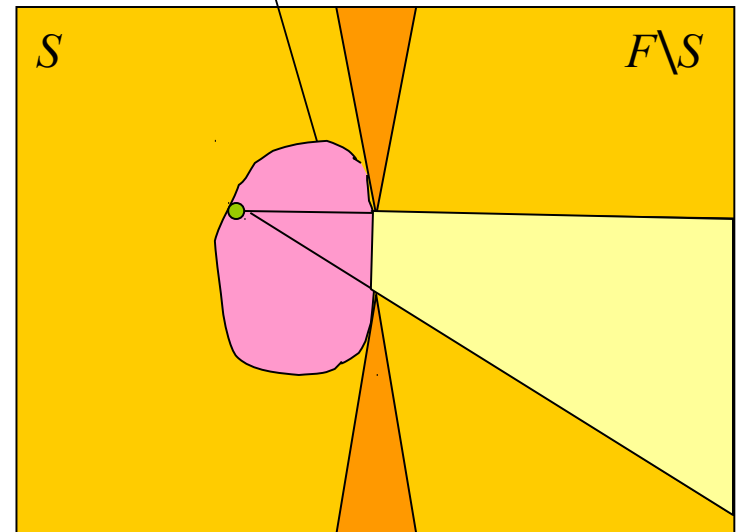
Definition: lookout of a subset S

- Subset of points in S that can see at least β fraction of $F \setminus S$, β is in $(0,1]$.

0.4-lookout of S

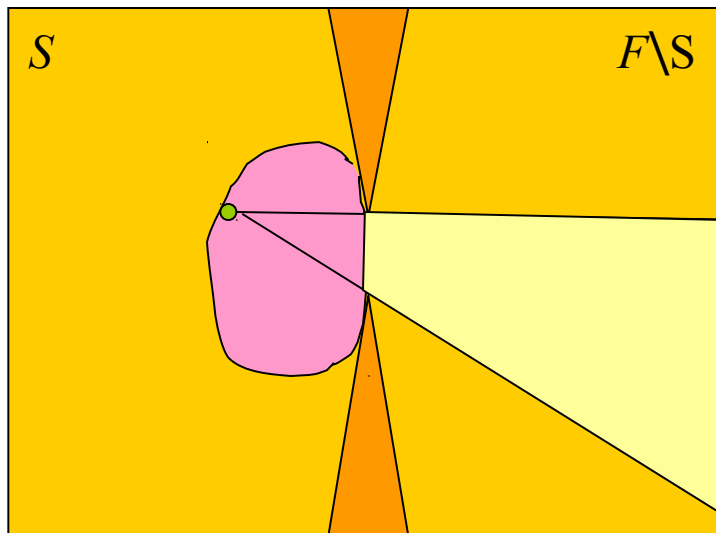


0.3-lookout of S



Definition: $(\epsilon, \alpha, \beta)$ -expansive

- The free space F is $(\epsilon, \alpha, \beta)$ -expansive if
 - Free space F is ϵ -good
 - For each subset S of F , its β -lookout is at least α fraction of S . ϵ, α, β are in $(0, 1]$



F is ϵ -good $\rightarrow \epsilon=0.5$

β -lookout $\rightarrow \beta=0.4$

$$\frac{\text{Volume}(\beta\text{-lookout})}{\text{Volume}(S)} \rightarrow \alpha=0.2$$

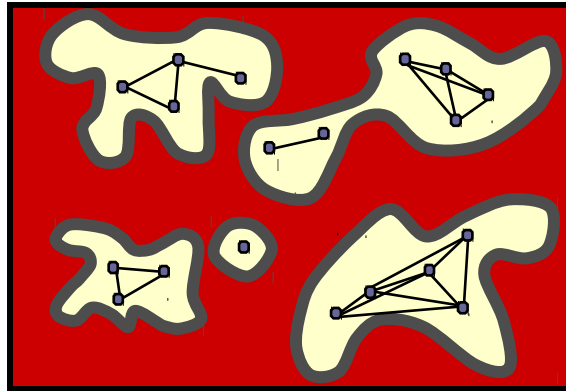
F is $(\epsilon, \alpha, \beta)$ -expansive,
where $\epsilon=0.5, \alpha=0.2, \beta=0.4$.

Why expansiveness?

- ϵ, α , and β measure the expansiveness of a free space.
- Bigger ϵ , α , and β \rightarrow lower cost of constructing a roadmap with good connectivity and coverage.

Uniform sampling

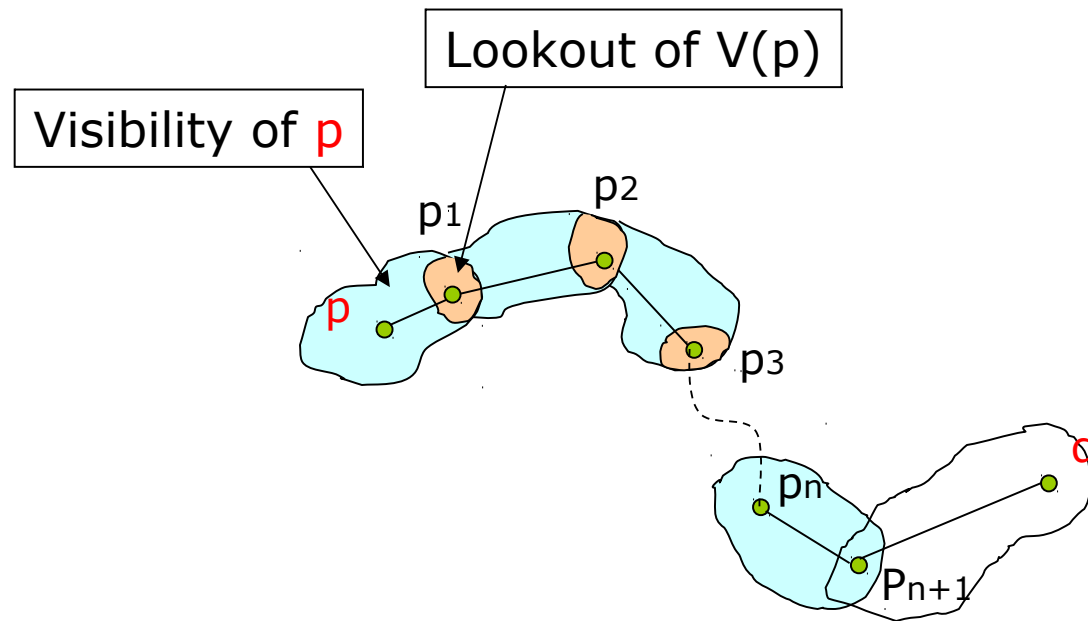
- All-pairs path planning



- **Theorem 1** : A roadmap of $\frac{16 \ln(1/\gamma)}{\epsilon \alpha} + \frac{6}{\beta}$

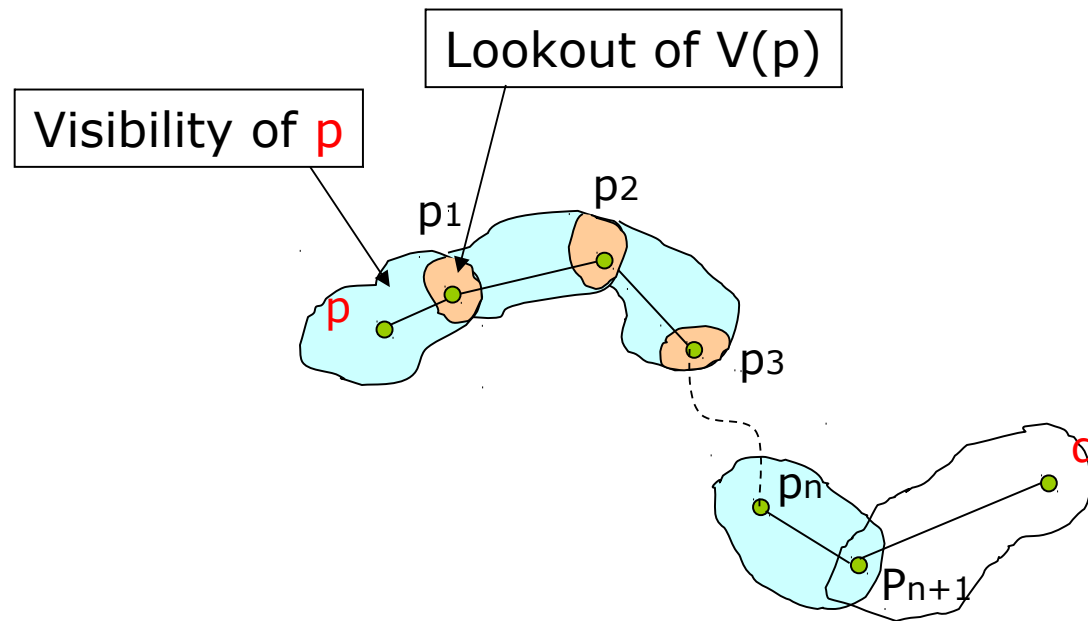
uniformly-sampled milestones has the correct connectivity with probability at least $1 - \gamma$.

Definition: Linking sequence



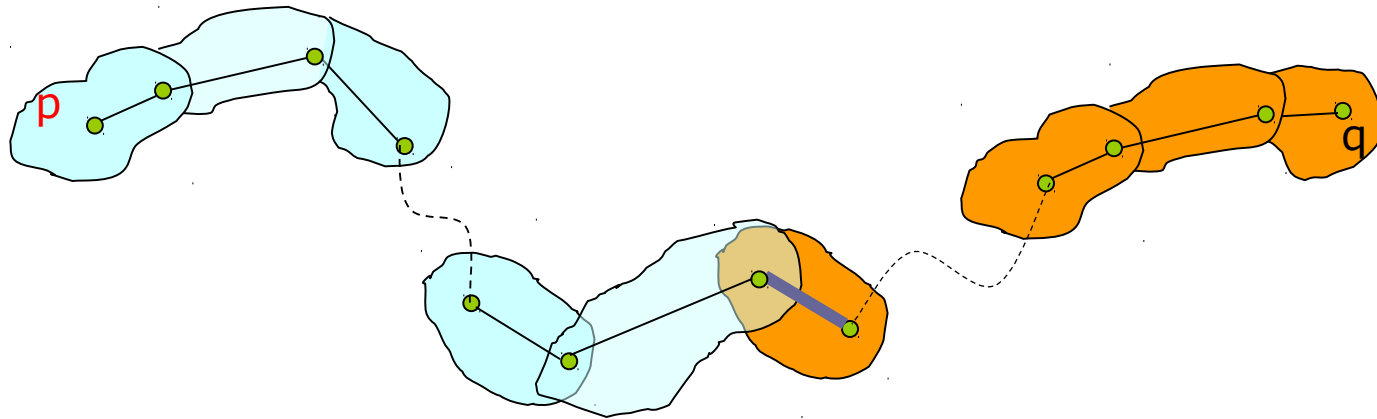
P_{n+1} is chosen from the lookout of the subset seen by p, p_1, \dots, p_n

Definition: Linking sequence

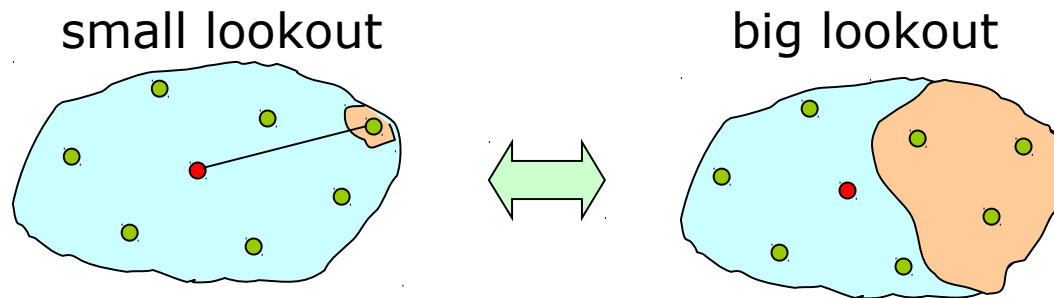
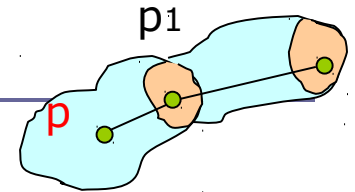


P_{n+1} is chosen from the lookout of the subset seen by p, p_1, \dots, p_n

Space occupied by linking sequences



Size of lookout set



A C-space with larger lookout set has **higher probability** of constructing a linking sequence.

Lemmas

- In an expansive space with large ε, α , and β , we can obtain a linking sequence that covers a large fraction of the free space, with high probability.

Theorem 1

- Probability of achieving good connectivity **increases exponentially** with the number of milestones (in an expansive space).
- If $(\epsilon, \alpha, \beta)$ **decreases** \rightarrow then need to **increase the number of milestones** (to maintain good connectivity)

Theorem 2

- Probability of achieving good coverage, **increases exponentially** with the number of milestones (in an expansive space).

Probabilistic completeness

In an expansive space, the probability that a PRM planner fails to find a path when one exists goes to 0 **exponentially** in the number of milestones (\sim running time).

[Hsu, Latombe, Motwani, 97]

Summary

□ Main result

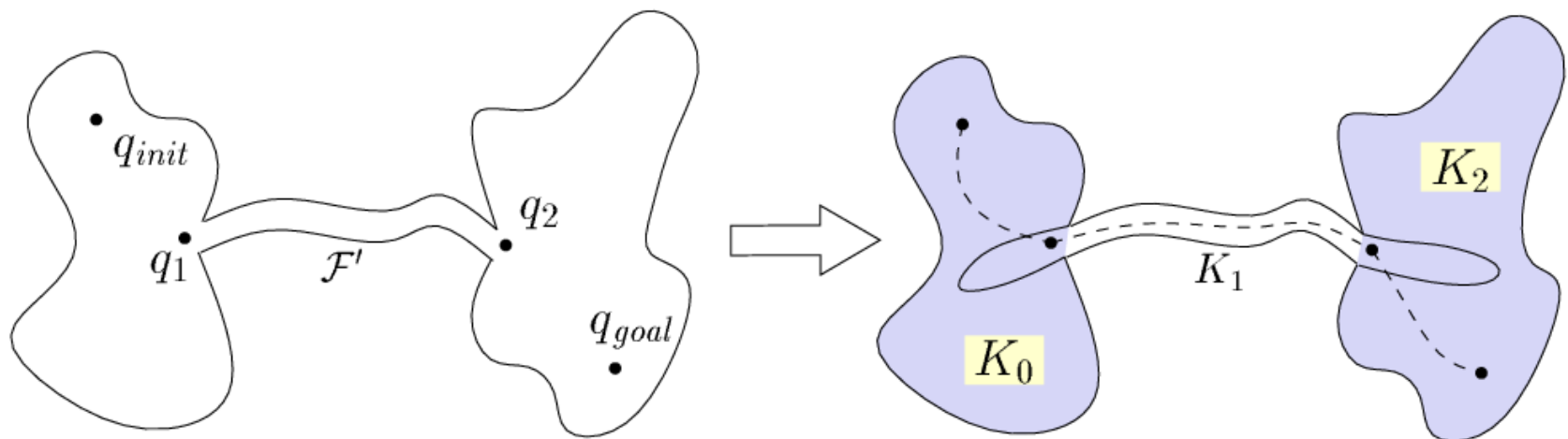
- If a C-space is expansive, then a roadmap can be constructed efficiently with good connectivity and coverage.

□ Limitation in practice

- It does not tell you when to stop growing the roadmap.
- A planner stops when either a path is found or max steps are reached.

Extensions

- Accelerate the planner by automatically generating intermediate configurations to decompose the free space into expansive components.



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- Use geometric transformations to increase the expansiveness of a free space, *e.g.*, widening narrow passages.
- Integrate the new planner with other planner for multiple-query path planning problems.

Questions?

Two tenets of PRM planning

- A relatively small number of milestones and local paths are sufficient to capture the connectivity of the free space.
 - Exponential convergence in expansive free space (probabilistic completeness)
- Checking sampled configurations and connections between samples for collision can be done efficiently.
 - Hierarchical collision checking