

### COS 495 – Lecture 20 Autonomous Robot Navigation

Instructor: Chris Clark Semester: Fall 2011

Figures courtesy of Siegwart & Nourbakhsh



#### **Control Structure**





# **Single Query PRMs: Outline**

- 1. Introduction
- 2. Algorithm Overview
- 3. Sampling strategies



# Motion Planning: Probabilistic Road Maps

- Single-Query PRMs (a.k.a. Rapidly Exploring Random Trees - RRTs)
  - Try to only sample a subspace of F that is relevant to the problem.
  - Probabilistically complete assuming C is *expansive* [Hsu et. al. 2000].
  - Very fast for many applications (allow for on-the-fly planning).



# Motion Planning: Probabilistic Road Maps

- Two approaches:
  - 1. Single Directional:
    - Grow a milestone tree from start configuration until the tree reaches the goal configuration
  - 2. Bi-Directional:
    - Grow two trees, one from the start configuration and one from the goal configuration, until the two trees meet.
    - Can't consider time in the configuration space



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Example:





















#### Example: Construct Path





#### Example: Construct Path





## Probabilistic Road Maps: Learning Phase

- Nomenclature
  - R=(N,E)RoadMapNSet of NodesESet of edgescConfigurationeedge



# Motion Planning: Probabilistic Road Maps

- Algorithm
  - 1. Add start configuration  $c_{start}$  to R(N,E)
  - 2. Loop
  - 3. Randomly Select New Node *c* to expand
  - 4. Randomly Generate new Node *c*' from *c*
  - 5. If edge e from c to c' is collision-free
  - 6. Add (c', e) to R
  - 7. If c' belongs to endgame region, return path
  - 8. Return if stopping criteria is met



# **Single Query PRMs: Outline**

- 1. Introduction
- 2. Algorithm Overview
- 3. Sampling strategies
  - Node Selection (step 3)
  - Node Generation (step 4)
  - Endgame Region (step 7)



## Motion Planning: PRM Node Selection

- One could pick the next node for expansion by picking from all nodes in the roadmap with equal probability.
  - This is easy to implement, but leads to poor expansion → Clustering
  - Method is to weight the random selection of nodes to expand, this can greatly affect the roadmap coverage of the configuration space.
  - Want to pick nodes with probability proportional to the inverse of node density.



### Motion Planning: PRM Node Selection

- Example:
  - Presented is a 2DOF configuration space where the initial node in the roadmap is located in the upper right corner.
  - After X iterations, the roadmap produced from an unweighted expansion has limited coverage.



Unweiahted







# Motion Planning: PRM Node Selection Technique 1

- The workspace was divided up into cells to form a grid [Kindel 2000].
  - Algorithm:
    - 1. Randomly pick an occupied cell from the grid.
    - 2. Randomly pick a milestone in that cell.





# Motion Planning: PRM Node Selection Technique 2

- Commonly used in Rapidly exploring Random Trees (RRTs) [Lavalle]
  - Algorithm:
    - 1. Randomly pick configuration  $c_{rand}$  from C.
    - 2. Find node c from R that is closest to node  $c_{rand}$
    - 3. Expand from c in the direction of  $c_{rand}$





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# Motion Planning: PRM Milestone Generation

- Use random control inputs to propagate robot from previous node c to new configuration c'
  - Algorithm:
    - 1. Randomly select controls u and  $\Delta t$
    - 2. Use known dynamics/kinematics equation *f* of robot to generate new configuration

$$c' = f(c, u, \Delta t)$$

3. If path from c to c' is collision-free, then add c' to R



## **Motion Planning: PRM Milestone Generation**

- **Example: Differential drive robot** 
  - 1. Randomly select controls  $\dot{\phi}_{left}$ ,  $\dot{\phi}_{right}$  and  $\Delta t$
  - 2. Propagate:
    - 1. Get  $\Delta s_{left}$  and  $\Delta s_{right}$  from  $\Delta s = v\Delta t$
    - 2. Calculate new state c' with:  $\left[\Delta s_r + \Delta s_i \cos\left(\Theta \pm \frac{\Delta s_r \Delta s_i}{1 \cos\left(\Theta \pm \frac{\Delta s_i}{1 \cos\left(\Theta \pm \frac{\Delta s_r \Delta s_i}{1 \cos\left$

$$c' = f(x, y, \theta, \Delta s_r, \Delta s_l) = \begin{bmatrix} x \\ y \\ \theta \end{bmatrix} + \begin{bmatrix} \frac{1}{2} \cos\left(\theta + \frac{1}{2b}\right) \\ \frac{\Delta s_r + \Delta s_l}{2} \sin\left(\theta + \frac{\Delta s_r - \Delta s_l}{2b}\right) \\ \frac{\Delta s_r - \Delta s_l}{b}$$

3. Use iterative search to check for collisions on path.



# Motion Planning: PRM Milestone Generation

- Example: Differential drive robot (cont')
  - Iterative Collision checking is simple but not always efficient:
  - Algorithm:
    - 1. Calculate distance *d* to nearest obstacle
    - 2. Propagate forward distance *d* along path from *c* to *c*'
    - 3. If *d* is too small, return *collision*
    - 4. If *c* reaches or surpasses *c*', return *collision-free*





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- We define the endgame region *E*, to be the set of configurations that have a *simple* connection to the goal configuration.
- For each planning problem, we can define a unique method of making *simple* connections.
- This method will inherently define *E*.





- Given the complexity of most configuration spaces, it is very difficult to model *E*.
- In practice, we develop a simple admissibility test to calculate if a configuration c' belongs to the E
- At every iteration of the algorithm, this test is used to determine if newly generated configurations are connected to the goal configuration.



- In defining *E*, we need two things for good performance:
  - 1. The region *E* should be large: this increases the chance that a newly generated milestone will belong to *E* and provide us a solution.
  - 2. The admissibility test to be as fast as possible. This test is conducted at every iteration of the algorithm and will greatly affect the algorithm running time.



- Several endgame definitions exist:
  - 1. The set of all configurations within some radius *r* of the goal configuration
  - 2. The set of all configurations that have "simple", collision-free connection with the goal configuration.
    - Example: Use circular arc for differential drive robots.





## Motion Planning: Probabilistic Road Maps

Video example