

# Probabilistic Robot Path Planning

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Readings:

- [Real-time randomized path planning for robot navigation](#), James Bruce and Manuela Veloso. In *Proceedings of IROS-2002*.

## Robot Motion Planning

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- A mobile robot needs to *navigate*:
  - Navigation is carrying out locomotion primitives to move between points
  - Navigation includes avoiding obstacles.
- We need to define:
  - The state – a model of the environment
  - The actions – a model of the robot's motion primitives

## Deterministic Path Planning

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- A\*
  - Discretize the state
  - Enumerate a set of actions
  - Search
    - Generate successors of states
    - Use admissible heuristic
  - Partially successful
- Extensions of A\*
  - D\*, etc

## Probabilistic Path Planning

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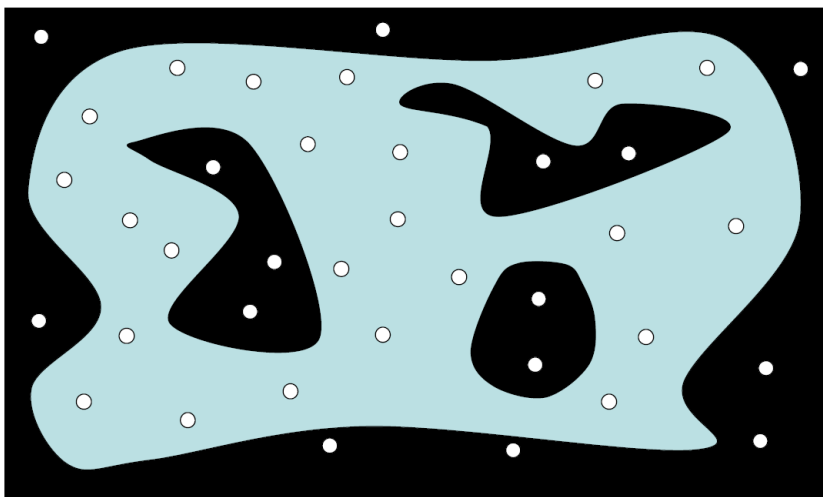
- Continuous state spaces
- Continuous actions
- PRM (Kavraki & many successors)
- RRT (Lavalle & many successors)
  - ERRT (Bruce & many other variations)

## PRM – Probabilistic Roadmap

- Separate planning into two stages
  - “Learning” Phase
    - random samples of free configurations (vertices)
    - Attempt to connect pairs of nearby vertices with a local planner
    - if a valid plan is found, add an edge to the graph
  - Query Phase
    - find local connections to graph from initial and goal positions

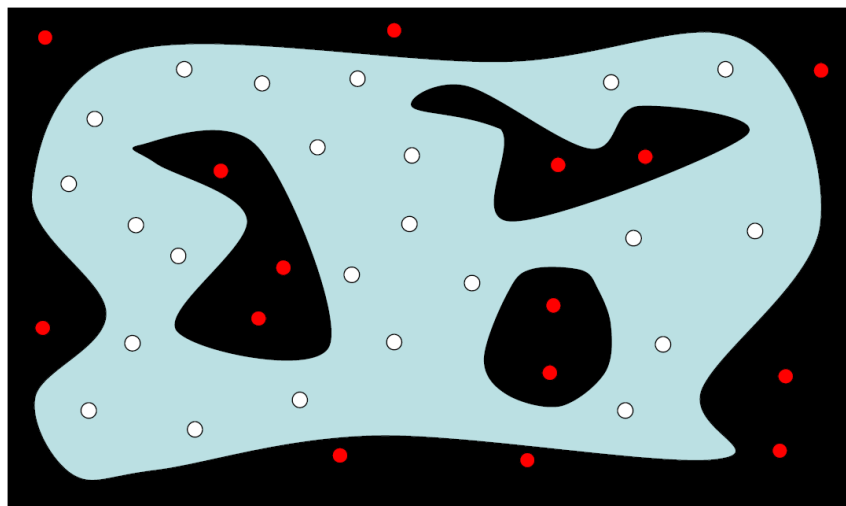
### Probabilistic Road Mapping

Sample random locations



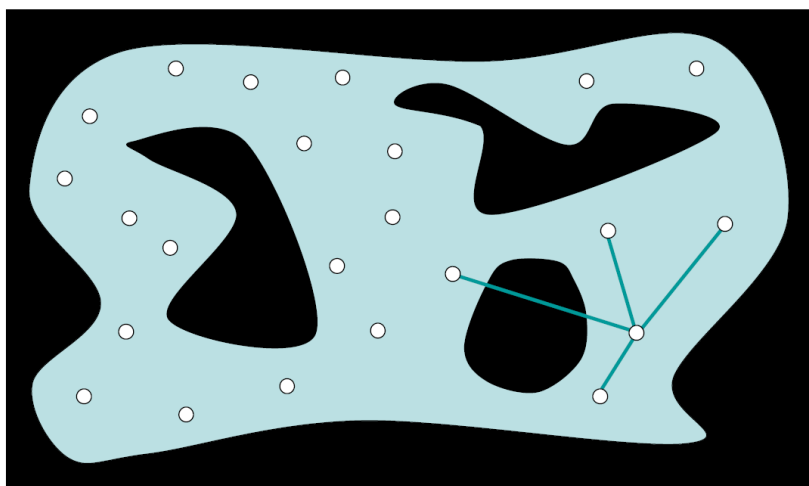
## Probabilistic Road Mapping

Remove the samples in the forbidden regions



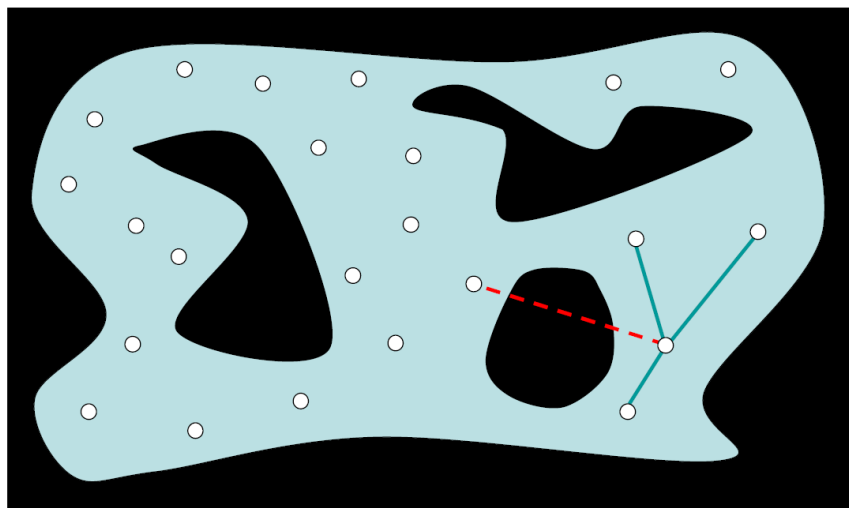
## Probabilistic Road Mapping

- Link each sample to its  $K$  nearest neighbors -



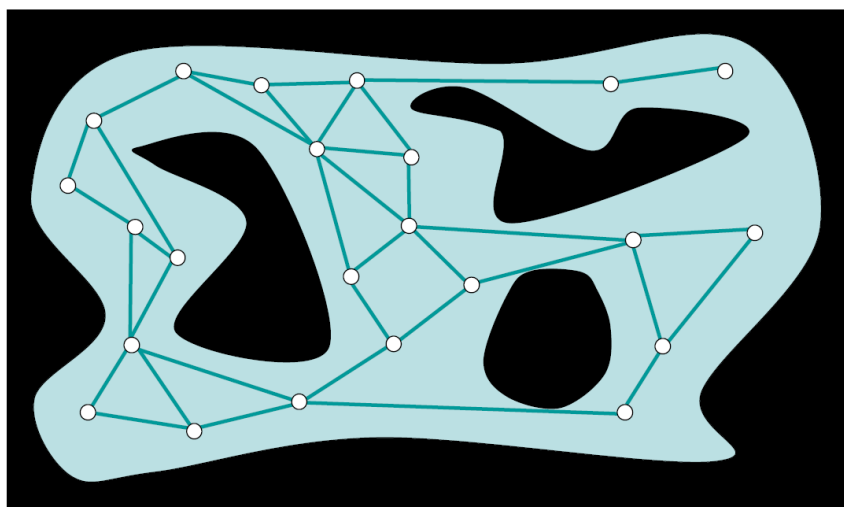
## Probabilistic Road Mapping

Remove the links that cross forbidden regions



## Probabilistic Road Mapping

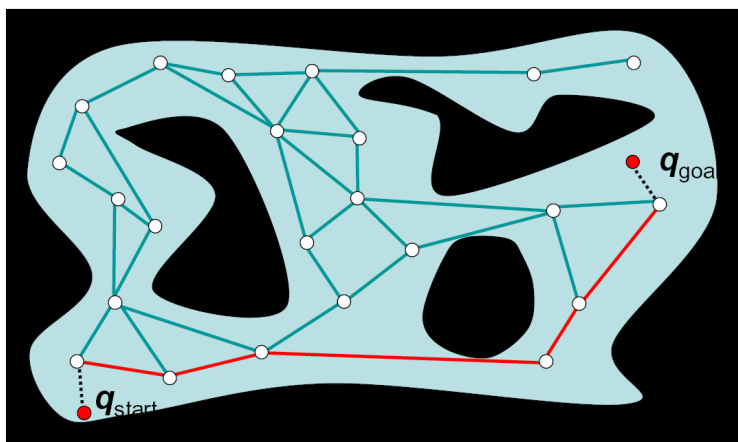
Remove the links that cross forbidden regions



The resulting graph is a *probabilistic roadmap (PRM)*

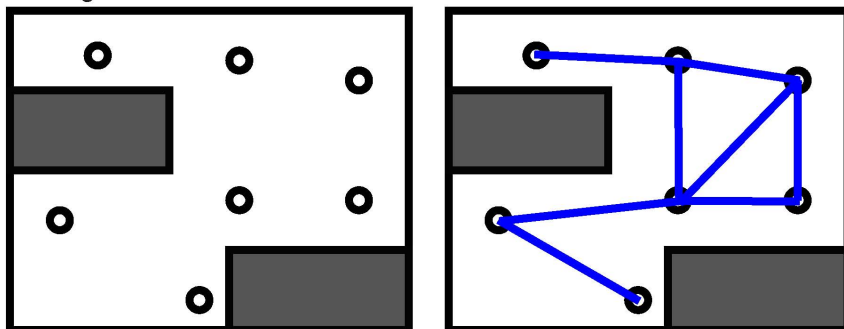
## Probabilistic Road Mapping

Link the start and goal to the PRM and search using A\*



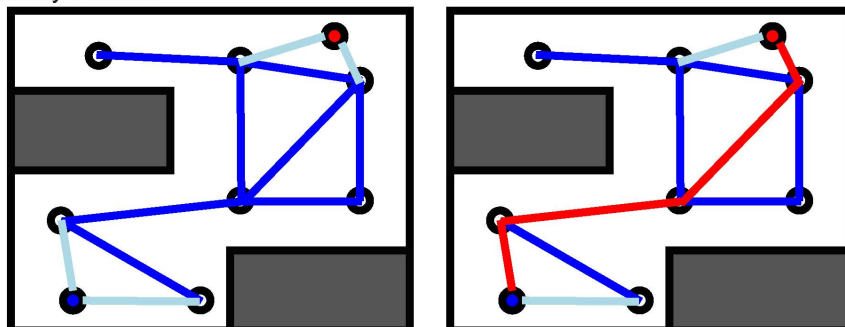
## PRM Example – Learning Phase

Learning Phase:



## PRM Example – Query Phase

Query Phase:



## PRM Discussion

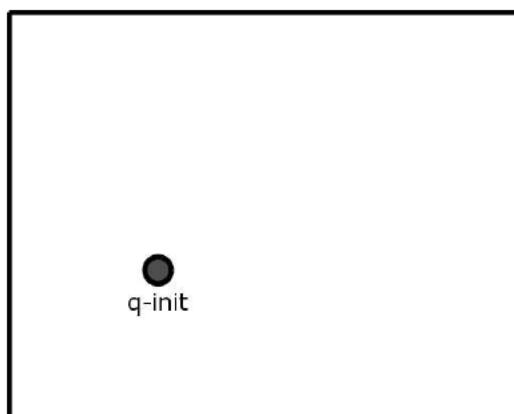
- Very interesting approach
  - Continuous spaces
- General learning phase
  - Not targeted at specific initial and goal states
- Not optimal path generated – probabilistically optimal
- How to pick sample configurations?
  - **Uniform** – simple, but may be slow
  - **Non-uniform** – (probabilistically) choose configurations that are in the neighborhood of nodes that are unconnected
    - Weight nodes by the “difficulty” of the region
- Efficient local planner to connect robot position to PRM samples

## Rapidly Exploring Random Trees

- RRT
  - Explore continuous spaces efficiently
    - No need for an artificial grid
  - Basic for probabilistically complete planner
- RRT uses random search

### Basic RRT Example

(1) Start with the initial state as the root of a tree

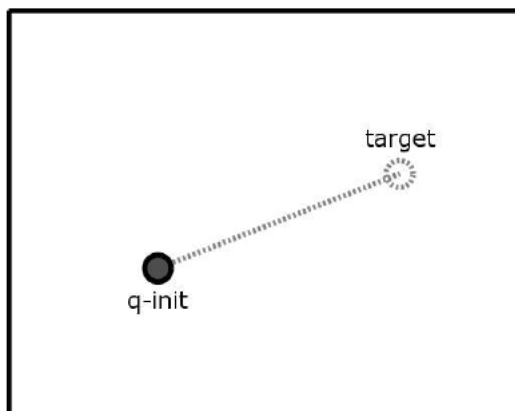




## Basic RRT

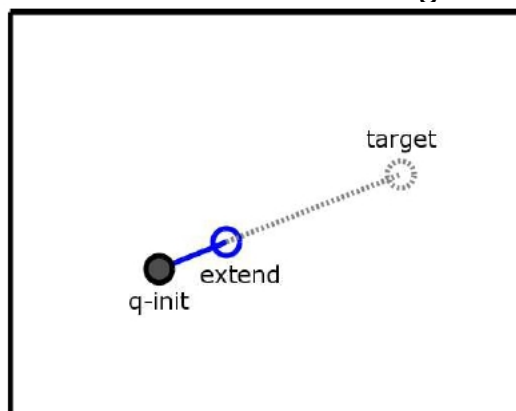
### ~~Just Search, No use of Goal~~

- (2) Pick a random state in the environment
- (3) Find the closest node in the tree

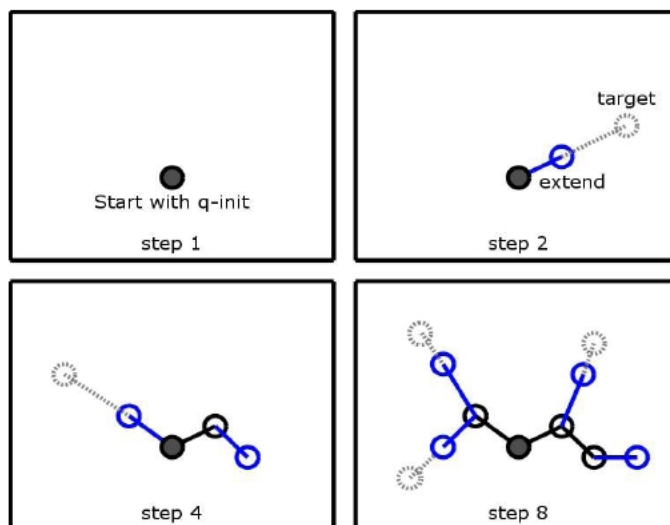


## Basic RRT Search

- (4) Extend that node toward the target



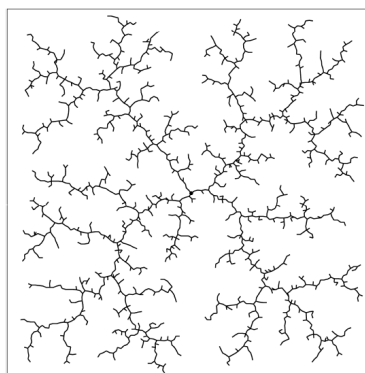
## Basic RRT Search Example



## RRT Basic Algorithm

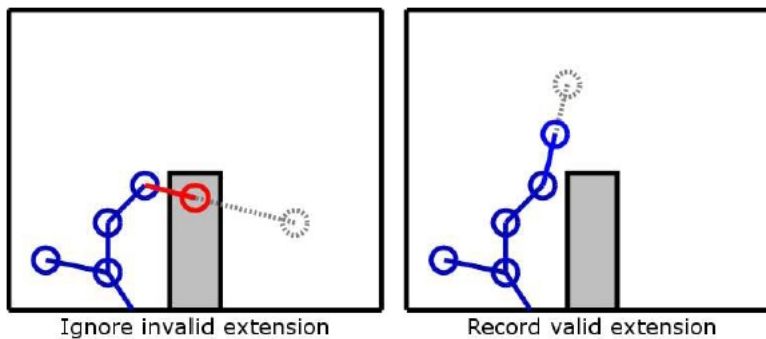
Pick a random point in the search space  
 Try to connect the current tree to that point  
 Continue until goal is reached (or stuck)

- Pick point  $p$  stochastically
- Find nearest node  $q$  in search tree
- Extend search from  $q$  in direction of point  $p$
- Continue until goal reached or limit on number of nodes
- If time (or goal not reached) can retry with different random seed



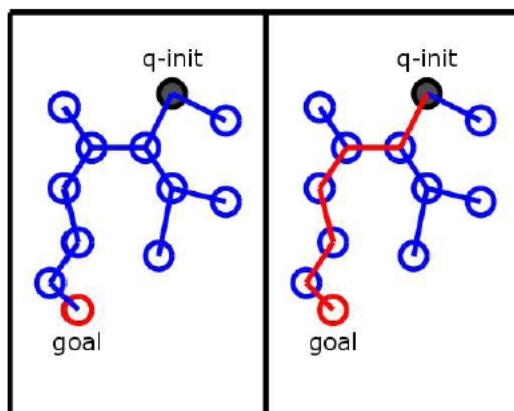
## RRT with Obstacles

- Ignore extensions which hit obstacles
- Resulting tree contains *only* valid paths



## RRT As a Planner

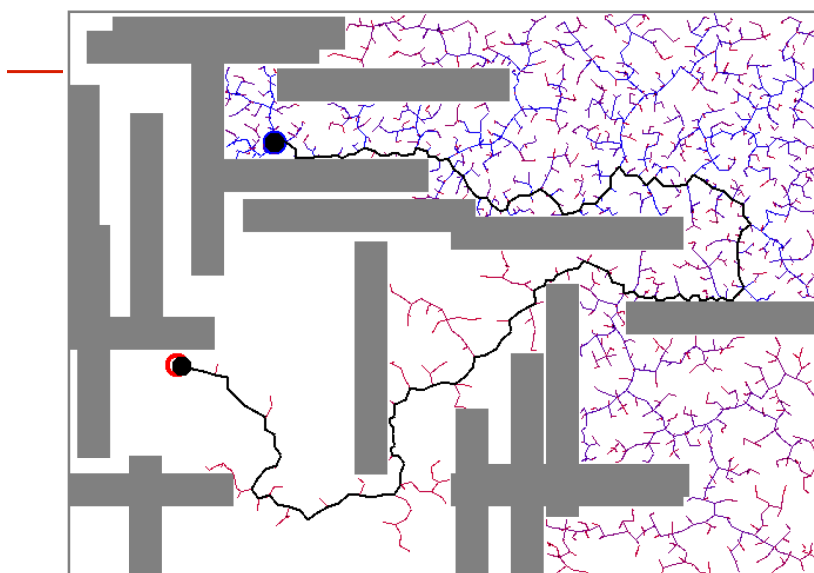
- Once a node of the tree is a *goal*, the plan is the path back up the tree



## RRT-GoalBias Algorithm

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- 1) Start with initial state as root of tree
- 2) Pick a random target state
  - o Goal configuration with probability  $p$
  - o Random configuration with probability  $1-p$
- 3) Find the closest node in the tree
- 4) Extend the closest node toward the target
- 5) Goto step 2



## ERRT – RRT with Replanning

(Bruce & Veloso 2002)

*Introduce past path(s) as a bias!*

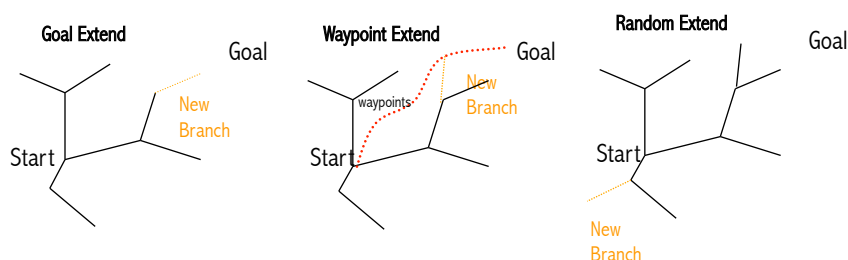
- 1) Start with initial state as root of tree
- 2) Pick a random target state
  - o Goal configuration with probability  $p$
  - o **Random item from waypoint cache with probability  $q$**
  - o Random configuration with probability  $1-q-p$
- 3) Find the closest node in the tree
- 4) Extend the closest node toward the target
- 5) Goto step 2

## ERRT: Replanning with Advice

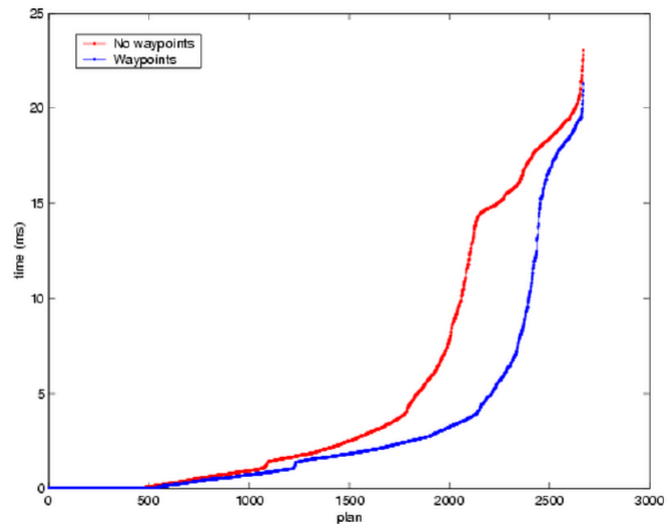
Probability  $p$  : Extend closest node in tree towards goal

Probability  $q$  : Extend closest node in tree towards random cache point

Probability  $1-p-q$  : Extend closest node towards a random point



## Waypoints [Results]



## Planning and Replanning

- Environments and planning
  - Value of  $p$ ?
- Dynamic environments
- When failure, what to do?

## Other RRT Variations

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- Use any heuristics to guide RRT (Urmson)
- Model uncertainty with particles (Melchior)
- Balanced growth sampling (Zickler)

## Summary

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- PRM
  - Sampling and search among sample nodes
- Planning with RRT
  - Extend towards random target, or towards goal
  - High  $p$  – few known obstacles
  - Low  $p$  – many known obstacles
- Replanning with ERRT
  - Extend towards random target, goal, or past plan
  - High  $q$  – small dynamics (no state change)
  - Low  $q$  – high dynamics (lots of state change)
  - ERRT – bias to use previous plan; but could be any other bias
- RRT and ERRT – probabilistic convergence