

Henrik I Christensen

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Path Planning

· A path is a continuous mapping in C

$$\pi:[0,L]\to S_{free}$$

- Where L is the length of the path
- The path is collision from if for all t

$$\pi(t) \in S_{free}$$

Query / problem definition

- A problem or query is
 - Given two states qo and qf
- Determine if there is a collision-free path between q₀ and q_f

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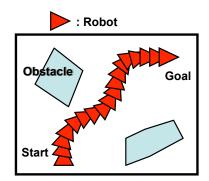
Path Planning

Given:

- World geometry
- Robot's geometry
- Start and goal configuration

Compute:

A collision-free, feasible path to the goal



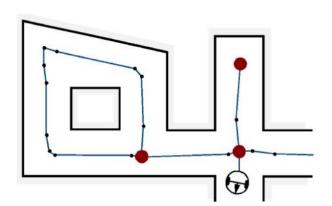


Source: F. Moss, Rice

Roadmaps

• Roadmaps / Graphs

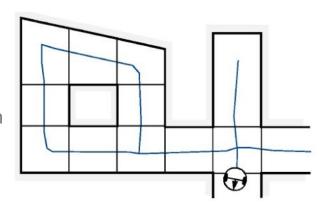
 How do we select the key nodes?



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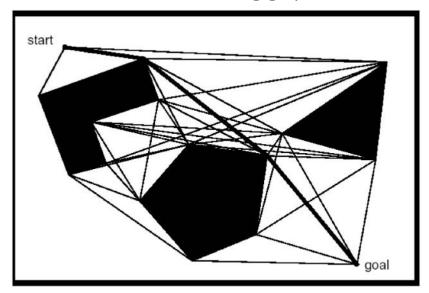
Space decomposition

- Partition
 - Free-space
 - Obstacles
- Generation of a tessellation



Visibility Graph

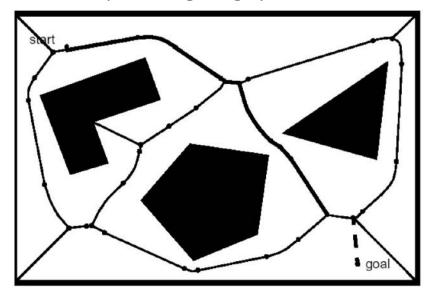
- Connect visible vertices in space
- Generate a search across the resulting graph



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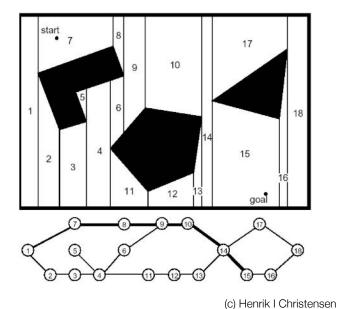
Voronoi Graph

- Compute maximum distance to boundary points
- Search for shortest path along the graph



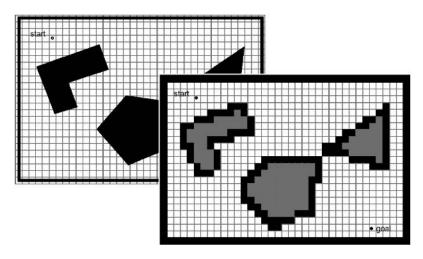
Cell decomposition

- Divide free space (Sfree) into single connected regions termed cells
- Generate connectivity graph
- · Local Goal and Start cells
- Search the graph
- Generation a motion strategy



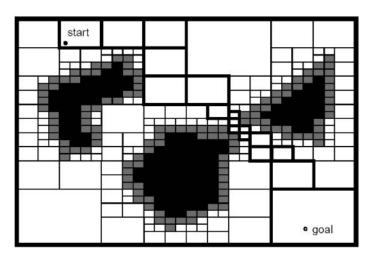
Approximate cell decomposition

- Easy to implement
- Use of standard methods for search such as wavefront & distance
- Widely used in simple environments



Adaptive cell decomposition

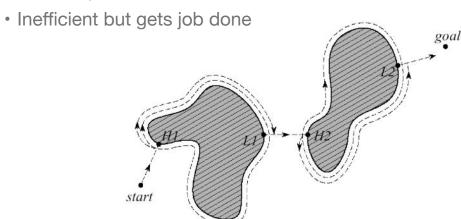
- Efficient representation of space
- Quad-trees are well-known from computational geometry
- Suited for sparse workspaces



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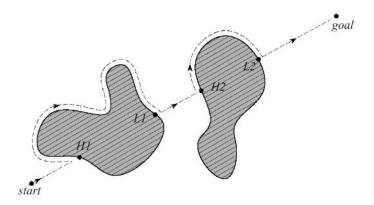
Planning basics

- The Bug family of algorithms [Lumelski]
- Simplest possible strategy
- Traverse obstacles
- · Leave a point of minimum distance



Bug-2 the obvious improvement

- Do traversal but leave at point to goal point
- Efficient in open spaces
- Mazes a challenge



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Potential fields

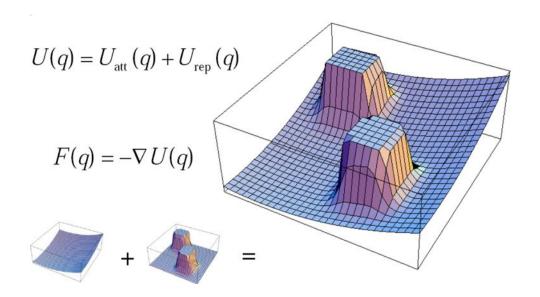
- · Consider the robot a particle in a potential field
- · Goal serves as an attractor
- Obstacles represents repellors
- · When the potential field is differentiable the force is

$$F(q) = -\nabla U(q)$$

specifies locally the direction of motion

- Potential fiedls can be represented by harmonics
 - · Superposition principle specifies
 - Goal dynamics can be represented by a potential field
 - · Each obstacle is a potential field
 - · The sum of the parts is still a potential field

Potential fields



Source: IROS tutorial, 2011

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Potential fields

· Potential fields are known to experience local minima

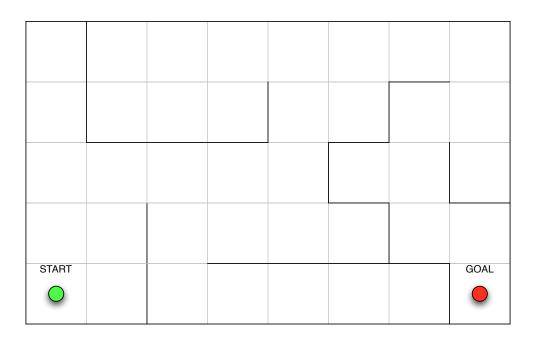


Wavefront propagation

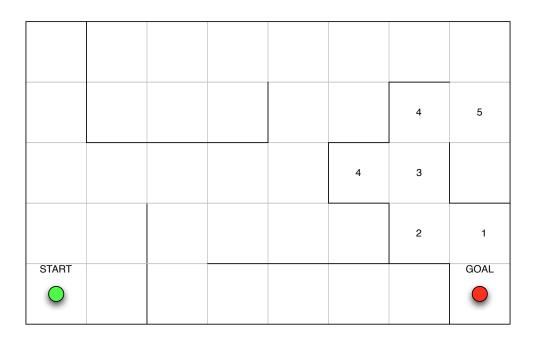
- Consider the world as a an homogenous grid
- · Cells are free or occupied / or with walls
- Search from start to goal
- Neighbor metrics can be used to define behavior

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Wavefront propagation



Wavefront propagation

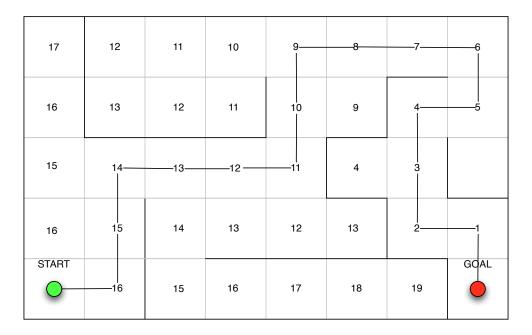


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Wavefront propagation

17	12	11	10	9	8	7	6
16	13	12	11	10	9	4	5
15	14	13	12	11	4	3	
16	15	14	13	12	13	2	1
START	16	15	16	17	18	19	GOAL

Wavefront propagation



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Graph search using A*

 A* is well known as a graph search heuristic based on estimated and actual cost

$$c(\vec{p}) = \alpha \ cc(\vec{s}, \vec{p}) + \beta \ gc(\vec{p}, \vec{g})$$

- · where
 - p is present position
 - s is the start position
 - g is the goal position
 - · cc is current cost
 - gc is an estimate of the cost of achieving the goal position
 - α, β represents weight factors

Basic tree-search

```
function Tree-Search( problem, fringe) returns a solution, or failure
  fringe ← Insert(Make-Node(Initial-State[problem]), fringe)
loop do
  if fringe is empty then return failure
    node ← Remove-Front(fringe)
  if Goal-Test[problem] applied to State(node) succeeds return node
  fringe ← InsertAll(Expand(node, problem), fringe)
```

The challenge is the design of the expand funtion

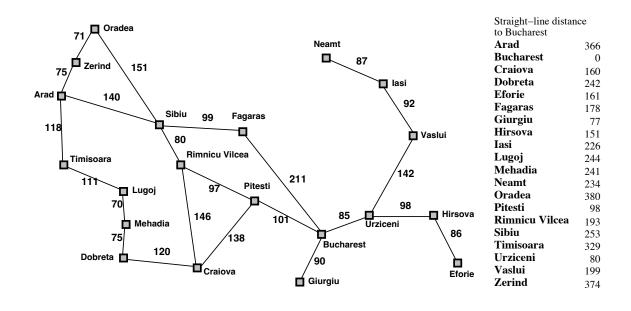
Source: Russell & Norvig, Artificial Intelligence

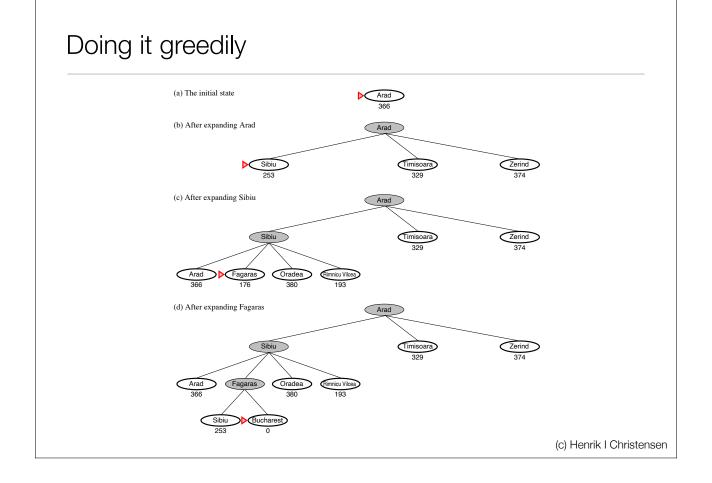
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Informed search strategies

- The ideal Best First Search
- Selection of an evaluation function f(n)
- Expand low-cost nodes before higher cost nodes
- Design a heuristic function h(n)
 - · Estimated cost of the cheapest path from n to goal

Example of navigation in maps



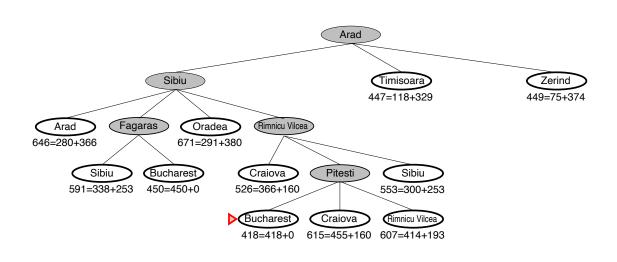


Properties of greedy search

- Completeness: might get stuck in loops
 - Repeated state check needed to break loops
- Time: O(bm) a good heuristic can improve performance
- Space O(bm) keep all nodes in memory
- Optimal? no greedy is not always optimal

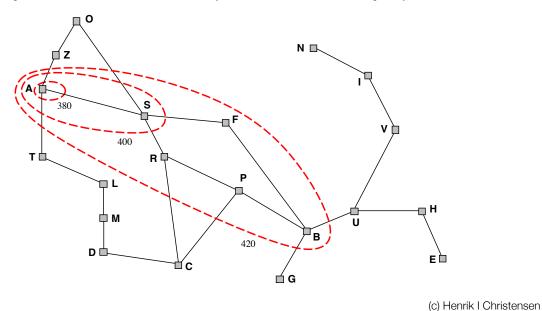
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A* search



A* optimality?

- · Increase nodes according to f value
- Gradually adds f contours to nodes (a la breadth first w. layers)



A* properties

- · Complete? Yes
- Time: exponential in h accuracy * h*(start)
- Space: all nodes in memory
- Optimal: Yes
- Variation of A*
 - Iterative deepening A* (IDA)
 - Recursive best first (RBDF)
 - Memory bounded A* (MA)
 - · Simple MA (SMA)

What are our options?

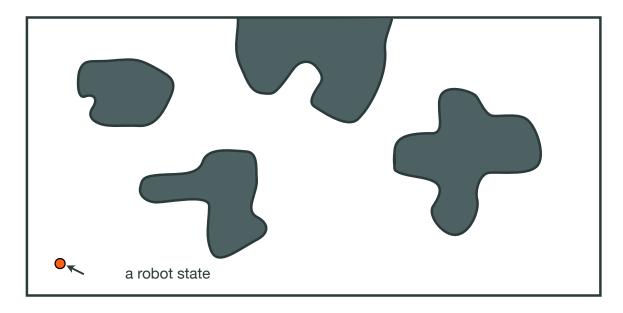
Method	Advantage	Disadvantage		
Exact	theoretically insightful	impractical		
Cell Decomposition	easy	does not scale		
Control-Based	online, very robust	requires good trajectory		
Potential Fields	online, easy	slow or fail		
Sampling-based	fast and effective	cannot recognize impossible query		

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Why randomized planners?

- The structure of the C-space can be highly complex
- The space can be high dimensional 6+

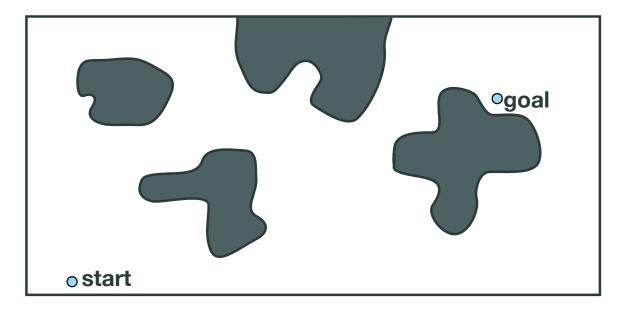
Point robot in 2-D



Source: L. Kavraki, RICE - Tutorial

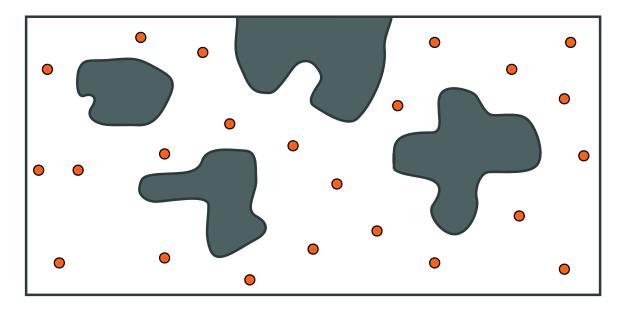
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Sample based tree planner



Source: L. Kavraki, RICE - Tutorial

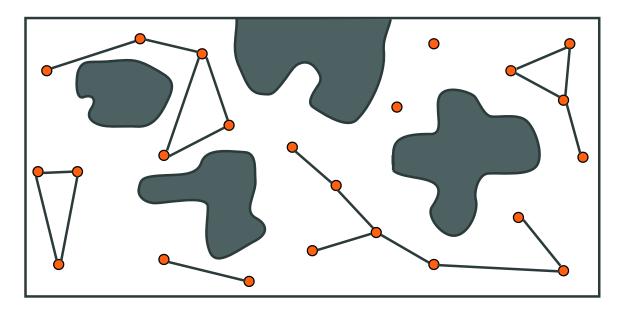
Probabilistic Roadmaps (PRM)



Source: L. Kavraki, RICE - Tutorial

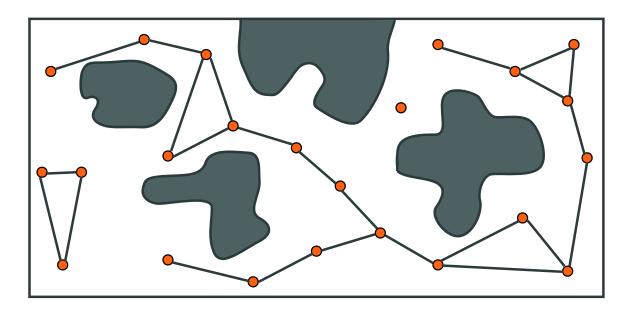
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Probabilistic Roadmaps (PRM)



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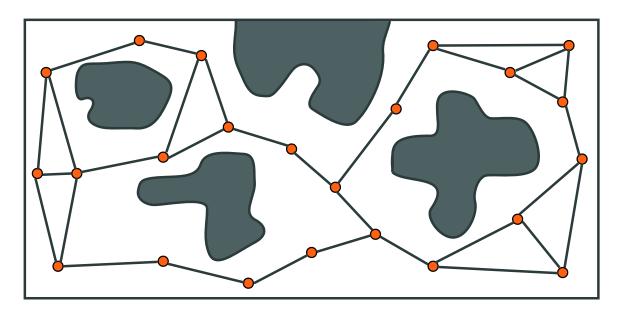
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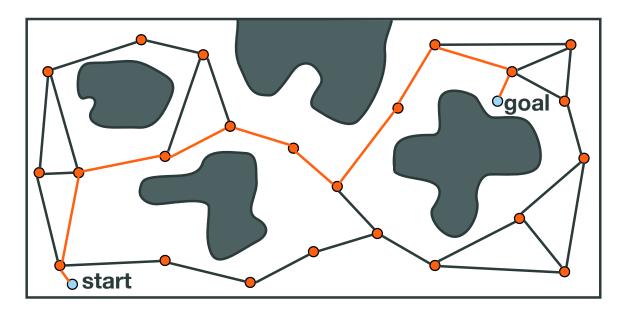
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Probabilistic Roadmaps (PRM)



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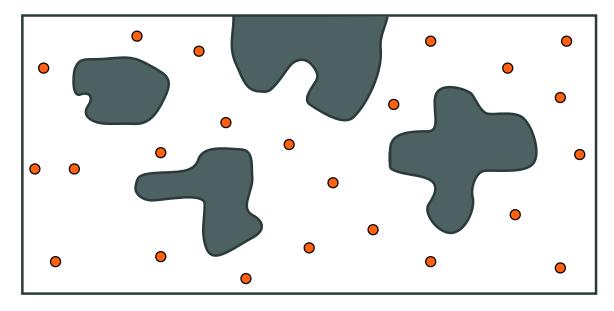
Answering queries



Source: L. Kavraki, RICE - Tutorial

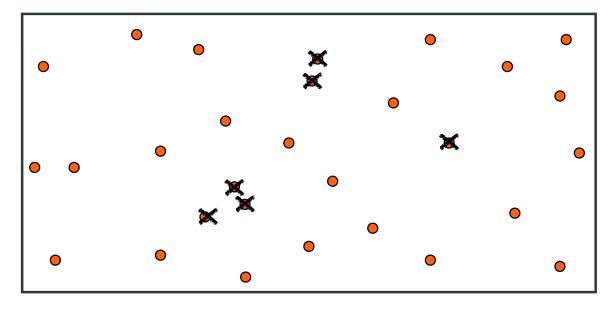
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Operations of a PRM



Source: L. Kavraki, RICE - Tutorial

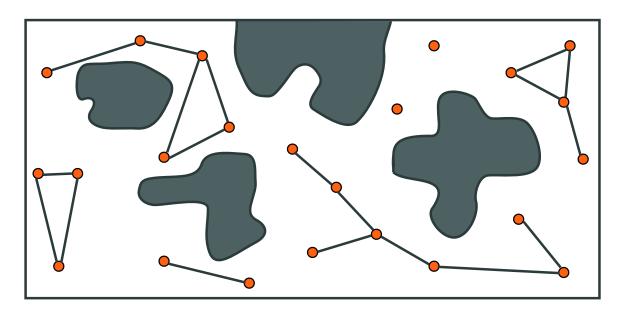
Operations of a PRM



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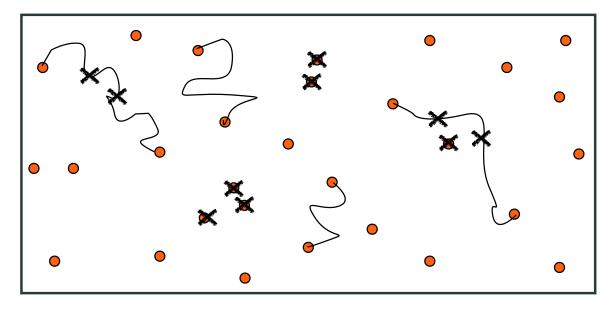
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Operations of a PRM



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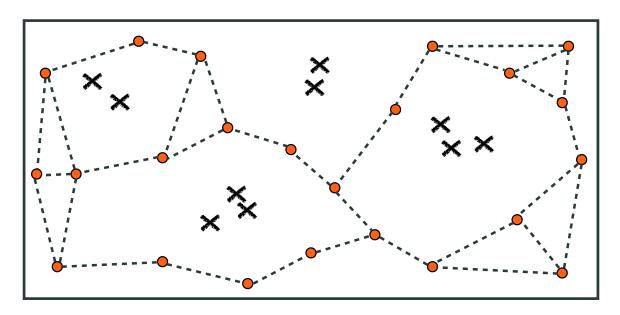
Operations of a PRM



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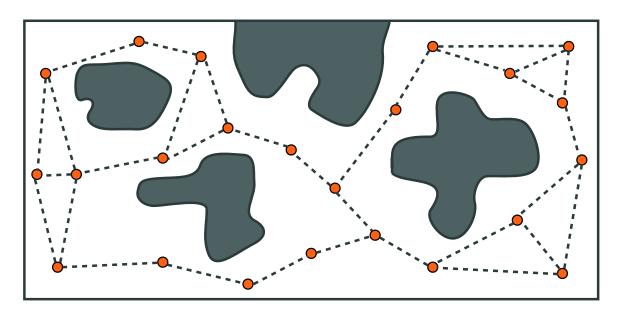
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Operations of a PRM



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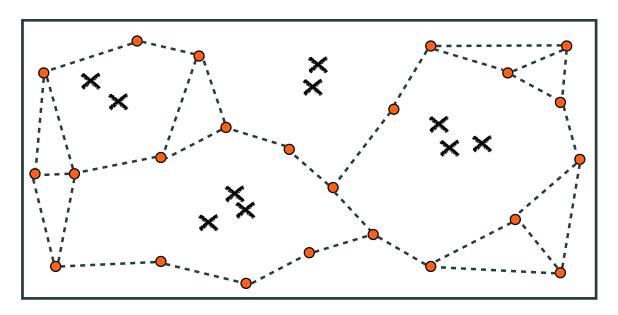
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Operations of a PRM



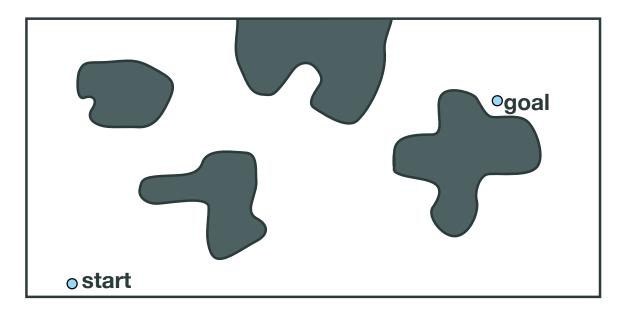
Source: L. Kavraki, RICE - Tutorial

Rapid Random Trees

- Could tree search be randomized to achieve some of the same functionality?
- There has been two recent approaches to randomized C-space search
 - Probabilistic Roadmaps (PRM)
 - Rapid Exploring Random Trees (RRT)

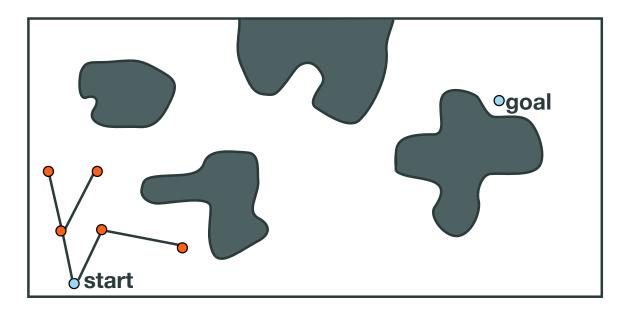
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Operations of a tree based planner



Source: L. Kavraki, RICE - Tutorial

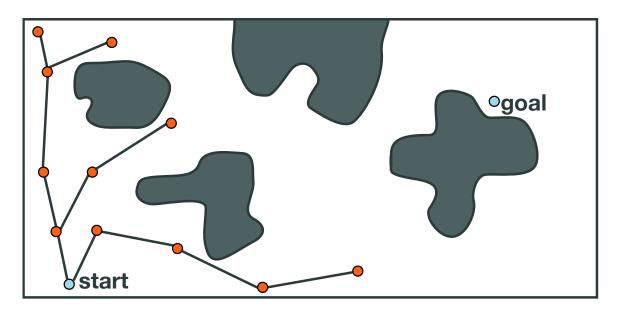
Sampling based tree planner



Source: L. Kavraki, RICE - Tutorial

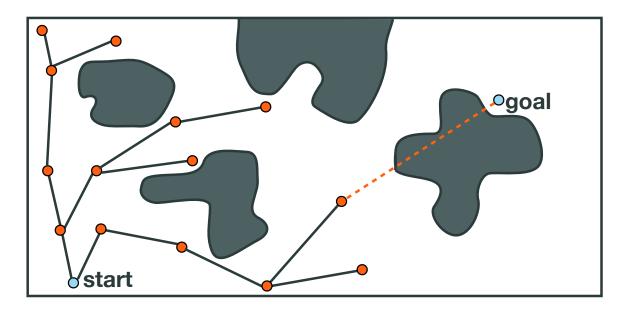
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Sampling based tree planner



Source: L. Kavraki, RICE - Tutorial

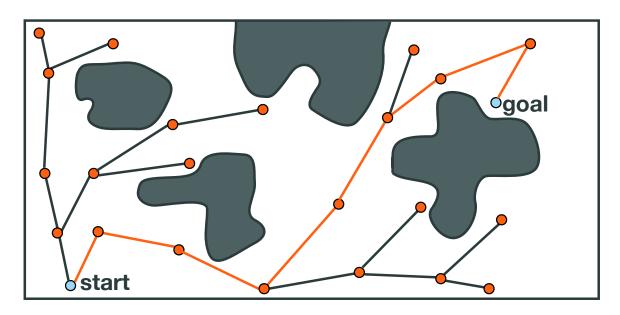
Sampling based tree planner



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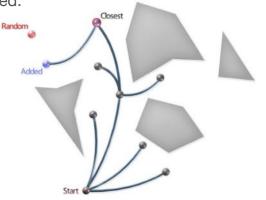
Sampling based tree planner



Source: L. Kavraki, RICE - Tutorial

Randomly Exploring Random Trees (RRT)

- Uses proximity query to guide construction (Voronoi Bias).
- Uses propagation instead of connection.
- Powerful heuristic for single-query planning.
- Bi-directional search can be implemented.



[Lavalle, Kuffer 1999, 2000]

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Planning

- There are a rich variety of planning methods
- · Consideration of the characteristics
 - Complexity of the configuration space?
 - Can domain constraints be imposed?
 - · Can we design deterministic search strategies?
 - What are memory requirements?
 - Do we need real-time response?
- · Repositories for planner benchmarking are emerging
- Great literature
 - Choset et al, Principles of Robot Motion, MIT Press
 - · Lavalle, Planning Algorithms, Cambridge University Press