

Computational Geometry

csci3250

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Approximate Path Planning

Geometric (combinatorial) path planning

- Approach

- **Compute free C-space geometrically**
 - compute C-obstacles as Minkowski sums and their union
- **Compute a partition of C-free**
 - trapezoidal partition
- **Compute a complete roadmap**

- **Geometric Roadmaps**

- Via trapezoidal partition or triangulation.
- Shortest Euclidian paths:
Visibility graphs
- Maximum clearance paths:
Voronoi diagram of obstacles

- Comments

- Complete
- Works beautifully in 2D and for some simple cases in 3D
- Worst-case bound for combinatorial complexity of C-objects in 3D is high
- Unfeasible for high #DOF
 - A complete planner in 3D runs in $O(2^{n^{\#DOF}})$

Approximate path planning

- Roadmaps
- Sampling
- Cell decomposition

Approximate path planning

- **Roadmaps**

- Compute a roadmap that captures connectivity of C-free

- **Sampling**

- When dimension of C-space is high it is hard to construct the C-obstacles
- However it is much easier to “sample” a certain placement and determine if it is in C-free or not.
 - this ends up being a collision test: would my robot , if placed at this position and in this configuration, intersect any obstacle?
- By sampling, we can determine what points are in C-free or not
- If the sampling is dense, it implicitly constructs a discretized version of C-free

Sampling

- You are not given the representation of C-free: Imagine being blindfolded in a maze
- Sampling: you walk around hitting your head on the walls
- Left long enough, after hitting many walls, you have a pretty good representation of the maze
- However the space is huge
 - e.g. DOF= 6: $1000 \times 1000 \times 1000 \times 360 \times 360 \times 360$
- So you need to be smart about it and chose the points you sample

Sampling and Roadmaps

- Instead of computing C-free explicitly, compute a roadmap that captures its connectivity
- Construct roadmap by sampling free space
- We can sample C-space
 - e.g. with a grid; or, randomly; or, ...
 - one shot, or incremental
 - ..
- If C-free is not sampled adequately, we may not detect paths that go through narrow passages
- Tradeoff: to capture all possible paths in C-free we want dense sampling, but this is infeasible because it's huge

Approximate path planning

- Goal: Approximate in some way C-free
- Approaches
 - Approximate the robot and/or obstacles by simple objects: (hierarchies of) bounding boxes or spheres
 - Cell decomposition method: Sample with uniform grid/quadtree/octree. This creates a huge roadmap which can be searched using AI techniques to direct the search towards a path
 - Potential method
 - Probabilistic roadmaps
 - RRT
 - any other approaches?
- Approaches are usually hybrid: combine ideas above

Use spheres and BB's

- Approximate the robot and/or obstacles by simple objects: (hierarchies of) bounding boxes or spheres
- Issues
 - cluttered scenes
 - complicated shapes

Grid roadmaps/cell decomposition

- Sample C-space with uniform grid/quadtree/octree.
- This essentially “pixelizes” the space
- May compute an occupancy grid
- Graph is implicit
 - given by grid topology: move +/-1 in each direction
- Can do it on one-shot (pre-compute it), or incrementally as needed
- This creates a huge roadmap which can be searched using AI techniques to direct the search towards a path

Potential field methods

- Idea [Latombe et al, 1992]
 - Define a potential field
 - Robot moves in the direction of steepest descent on potential function
- Ideally potential function has global minimum at the goal, has no local minima, and is very large around obstacles
- Algorithm outline:
 - place a regular grid over C-space
 - search over the grid with potential function as heuristic
- Pro: Nice framework that can be adapted to any specific scene
- Con: can get stuck in local minima
 - Potential functions that are minima-free are known, but expensive to compute
- RPP (Randomized path planner) is based on potential functions
 - escapes local minima by executing random walks
 - successfully used to
 - performs riveting ops on plane fuselages
 - plan disassembly ops for maintenance of aircraft engines

Probabilistic Roadmaps

- Latombe, Overmars et al , 1996
- Approach
 - **Construction phase:**
 - Construct roadmap representing C-free by repeatedly sampling and connecting
 - **Query phase:**
 - Use roadmap to find path between any two points

Probabilistic Roadmaps

- Constructing the roadmap
 - Start with uniform sampling of points in C-free and try to connect them
 - two points are connected by an edge if a simple quick planner can find a path between them
 - This will create a set of trees
 - Augment roadmap by selecting additional sample points in areas that are estimated to be “difficult”
- Comments
 - Roadmap adjusts to the density of free space and is more connected around the obstacles
 - Size of roadmap can be adjusted as needed
 - More time spent in the “learning” phase ==> better roadmap
 - Roadmap stored as set of tree for space efficiency
 - trees encode connectivity, cycles don't change it. Additional edges are useful for shortest paths, but not for completeness

Probabilistic Roadmaps

```
(1)   $N \leftarrow \emptyset$ 
(2)   $E \leftarrow \emptyset$ 
(3)  loop
(4)     $c \leftarrow$  a randomly chosen free
        configuration
(5)     $N_c \leftarrow$  a set of candidate neighbors
        of  $c$  chosen from  $N$ 
(6)     $N \leftarrow N \cup \{c\}$ 
(7)    for all  $n \in N_c$ , in order of
        increasing  $D(c,n)$  do
(8)      if  $\neg \text{same\_connected\_component}(c,n)$ 
         $\wedge \Delta(c,n)$  then
(9)         $E \leftarrow E \cup \{(c,n)\}$ 
(10)      update  $R$ 's connected
        components
```

- Components
 - sampling C-free
 - the local planner
 - selecting the neighbors
 - the expansion step and the heuristical measure of difficulty of a node

Probabilistic Roadmaps

Comments

- One of the leading motion planning technique
- Efficient, easy to implement, applicable to many types of scenes
- Embraced by many groups, many variants of PRM's, used in many type of scenes.
- Not clear which technique better in which circumstances