Lecture 17: *Motion Planning I*

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1 Complete Motion Planning

1.1 Element of Motion plan

Motion plan generally takes the elements followings into the considerations.

- **Plans** {Discrete, Continous, Hibrid}
  This category indicates whether a certain plan covers a discrete number space or a continuous one. For instance, grasping an object in a continuous space while make a decision among discrete classes would belong to Hybrid plans.
- **Environment** {Immovable, Movable, Moving}
  This is the environment where an agent moves. The example of Immovable environment can be fixed obstacles such as walls while movable ones are Movable objects like apples, chairs. Moving objects are literally the non-fixed object on the fields such as animals, humans.
- **Interaction** {None-Contact, Contact}
  An agent interacts with the environment, which should be considered during planning. None-Contact interaction is the interaction where the agent tries to avoid a collision while Contact embraces situations such as grasping, pushing, or pulling an object.
- **Uncertainty** {Node, Known, Bounded, Unknown}
  'Node' refers to given configuration spaces q while 'Known' does to the position according to the configurations, which is p(q). 'Bounded' refers to restrict conditions that bound our configurations (ex, q must belong to Q). 'Unknown' indicates unknown configuration that we need to figure out. Usually, the unknown configuration is a semi solution or solution itself of entire planning problems.
- **Robot Motion** {Kinematic, Dynamics}
  Kinematic setting describes a robot motion with configuration q while Dynamics does with configuration and its derivative q' that is, the velocity of q.

1.2   Problem Formulation(piano mover's problem)

Figure 1.2 describes the elements of Piano mover's problems with which we will usually deal in motion panning. The criteria settings of the problems are the followings:
- A world with R^2 or R^3 spaces.
- Obstacles occupies a space Cobs (Q_{obs}) in the world
- A robot and its configuration space $C(Q)$
- Where $Q_{free} = Q \setminus Q_{obs}$, qs is in Q_{free} at the initial configurations; and so is qg. Q_{free} is the configuration spaces that are not occupied with obstacles.
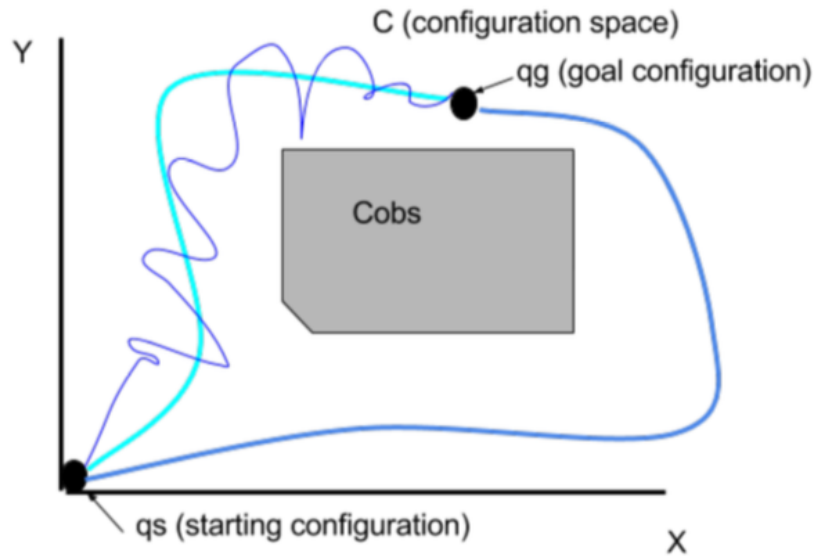
Figure 1: piano movers problem

For short, our goal as to that problem is to come up with a function that executes the mapping $T : [0,1] \rightarrow Q_{free}$ where $qs = T(0)$ and $qg = T(1)$. In other word, this is finding a path specifying configurations corresponding to each time steps between 0 and 1.

1.3   Cell Decompose

Considering the contents above, it is obvious that how to represent and know free space is a crucial issue to solve piano mover's problems. Cell Decompose is a category of a method to represent the free space where the free spaces are the set of 'cells' into which the free spaces are separated. Trapezoidal decomposition algorithm is one of them. Each node corresponds to a cell and an edge connects nodes of adjacent cells forming

graphs. After the cell decomposition, path planning with a cell decomposition is usually done in two steps: first, the planner determines the cells that contain the start and goal, respectively, and then the planner searches for a path within the adjacency graph. Note that the adjacency graph could serve as a roadmap of the free space as well. If there is one ore more path that connects start and goal, that means there is a path among the road map; thus, this solution is roadmap based method as well where it is guaranteed to derive a solution from a roadmap or at the very least know there is no valid route. Yet, this method can only be applied to simple spaces.
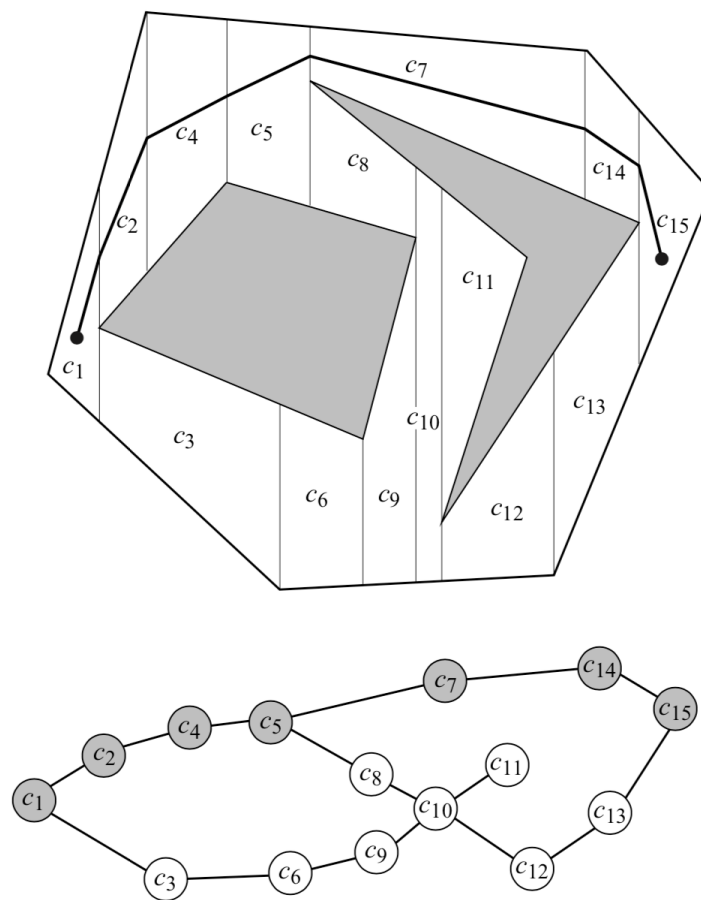
1.3.1   Algorithm

Figure 2: cell decompose

Fig 2 The resulting paths in the adjacency graph and free space. The straight lines up

and down are called 'boundary' while the separated areas (cells) are 'trapajoid'.

We assume all the vertices and their Euclidean positions are known. Then for each vertex, do the following steps.

- Draw line heading to upper and bottom from the vertex until it meets an intersection.
- Sample a point in every trapezoid and its boundary; these points will be saved as nodes of a graph. Note that every boundary line must exclude an intersection.

After doing that, find the trapezoid where our start and goal belongs to; and connect all the nodes forming a graph. Then, one can easily find a path by using the graph search algorithms such as DFS, or BFS.

1.4   Visibility Graph

The defining characteristics of a visibility map are that its nodes share an edge if they are within line of sight of each other, and that all points in the robot's free space are within line of sight of at least one node on the visibility map. This second statement implies that visibility maps, by definition, possess the properties of accessibility and departability. Connectivity must then be explicitly proved for each map for the structure to be a roadmap. In this section, we consider the simplest visibility map, called the visibility graph

1.4.1   Visibility Graph Definition

The standard visibility graph is defined in a two-dimensional polygonal configuration space (figure 2.1). The nodes v_i of the visibility graph include the start location, the goal location, and all the vertices of the configuration space obstacles. The graph edges $e_i j$ are straight-line segments that connect two line-of-sight nodes v_i and v_j, i.e.,

$$e_i j \neq \varnothing \iff sv_i + (1-s)v_i \in cl(Q_{free}) \forall s \in [0,1]$$

Note that we are embedding the nodes and edges in the free space and that edges of the polygonal obstacles also serve as edges in the visibility graph.

By definition, the visibility graph has the properties of accessibility and departability. We leave it to the reader as an exercise to prove the visibility graph is connected in a connected component of free space. Using the standard two-norm (Euclidean distance), the visibility graph can be searched for the shortest path . The visibility graph can be defined for a three dimensional configuration space populated with polyhedral obstacles, but it does not necessarily contain the shortest paths in such a space.

Unfortunately, the visibility graph has many needless edges. The use of supporting and separating lines can reduce the number of edges. A supporting line is tangent
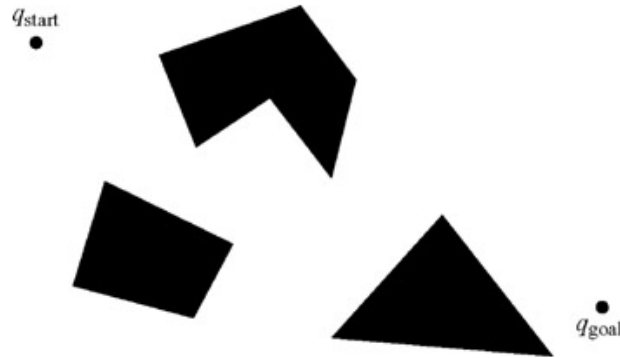
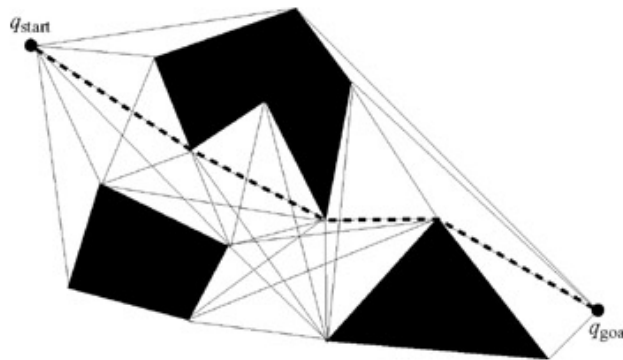Figure 3: Polygonal configuration space with a start and goal

Figure 4: The thin solid lines delineate the edges of the visibility graph for the three obstacles represented as filled polygons. The thick dotted line represents the shortest path between the start and goal

to two obstacles such that both obstacles lie on the same side of the line. For nonsmooth obstacles, such as polygons, a supporting line l can be tangent at a vertex v_i if $\beta_\epsilon(v_i) \bigcap l \bigcap QO_i = v_i$. A separating line is tangent to two obstacles such that the obstacles lie on opposite sides of the line.

The reduced visibility graph is soley constructed from supporting and separating lines. In other words, all edges of the original visibility graph that do not lie on a supporting or separating line are removed.
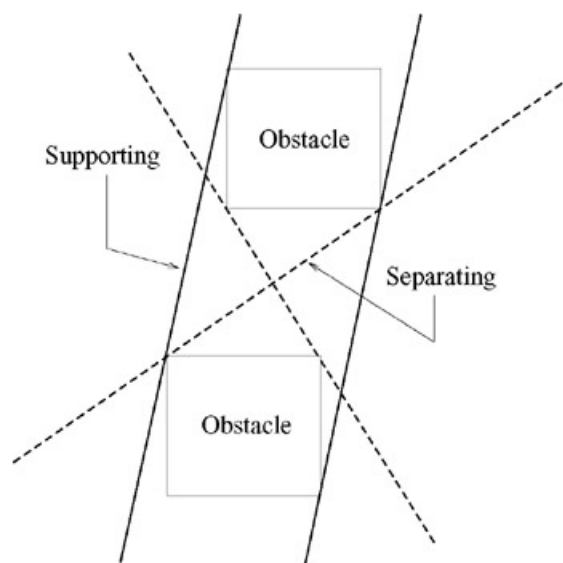
Figure 5: Supporting and separating line segments. Note that for polygonal obstacles, we use a nonsmooth notion of tangency

1.4.2 Visibility Graph Construction

Let $V = v_1, ..., v_n$ be the set of vertices of the polygons in the configuration space as well as the start and goal configurations. To construct the visibility graph, for each be the set of vertices of the polygons in the configuration space as well as the start and goal configurations. To construct the visibility graph, for each $v \in V$ we must determine which other vertices are visible to v.The most obvious wayto make this determination is to test all line segments $vv_i, v \neq v_i$to see if they intersect an edge of any polygon. For a particular vv_i, there are $O(n)$ intersections to check because there are $O(n)$edges from the obstacles. Now, there are $O(n)$ potential segments emanating from v,so for a particular v, there are $O(n_2)$ tests to determine which vertices are indeed visible from v, This must be done for all $v \in V$ and thus the construction of the visibility graph would have complexity $O(n_3)$

There is a more efficient way to compute the set of vertices that are visible from v. Imagine a rotating beam of light emanating from a lighthouse beacon. At any moment, the beam illuminates the object that is closest to the lighthouse. Furthermore, as the beam rotates, the obstacle that is illuminated changes only at a finite number of orientations of the beam. If the obstacles in the space are polygons, these orientations occur when the beam is incident on a vertex of some polygon. This insight motivates a class of algorithms known in the computational geometry literature as plane sweep algorithms.

A plane sweep algorithm solves a problem by sweeping a line, called the sweep line,
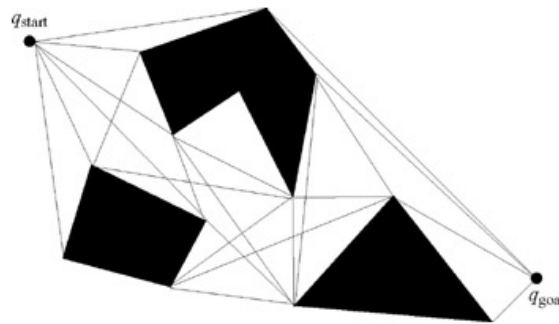
Figure 6: Reduced visibility graph

across the plane, pausing at each of the vertices of the obstacles. At each vertex, the algorithm updates a partial solution to the problem. Plane sweep algorithms are used to efficiently compute the intersections of a set of line segments in the plane, to compute intersections of polygons, and to solve many other computational geometry problems. For the problem of computing the set of vertices visible from v, we will let the sweep line, l, be a half-line emanating from v,and we will use a rotational sweep, rotating l from 0 to 2π. The key to this algorithm is to incrementally maintain the set of edges that intersect l , sorted in order of increasing distance from v. If a vertex v_i is visible to v, then it should be added to the visibility graph.It is straightforward to determine if v_i is visible to v (see figure 7). Let S be the sorted list of edges that intersects the half-line emanating from v; the set S is incrementally constructed as the algorithm runs. If the line segment vv_i does not intersect the closest edge in S, and if$l$ does not lie between the two edges incident on v,(the sweep line does not intersect the interior of the obstacle at v), then the v_i is visible to v.

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Algorithm 5: Rotational Plane Sweep Algorithm

Input: A set of vertices {vᵢ} (whose edges do not intersect) and a vertex v
Output: A subset of vertices from {vᵢ} that are within line of sight of v
1: For each vertex vᵢ, calculate αᵢ, the angle from the horizontal axis to the line segment
vvᵢ.
2: Create the vertex list ε, containing the αᵢ 's sorted in increasing order.
3: Create the active list S, containing the sorted list of edges that intersect the horizontal
half-line emanating from v.
4: for all αᵢ do
5:    if vᵢ is visible to v then
6:       Add the edge (v, vᵢ )to the visibility graph.
7:    end if
8:    if vᵢ is the beginning of an edge, E, not in S then
9:       Insert the E into S.
10:   end if
11:   if vᵢ is the end of an edge in S then
12:      Delete the edge from S.
13:   end if
14: end for
```

Figure 7: Rotational Plane Sweep Algorithm

2 Sampling-based Motion Planning

Sampling-based motion planning is not complete: not guarantee to find a solution when one exists. However, as the number of samples goes to infinity, there is a strong guarantee that it can find a solution.

There are 2 types of sample-based planning algorithms:

Multiple Query Algorithm
- Roadmap is built beforehand. Then it can be used multiple times for different queries.
- Query is fast since roadmap is already generated.
- Keeping roadmap all the time is expensive.
- Can only deal with static environment - has to rebuild the roadmap whenever anything changed in environment.

Single Query Algorithm
- No roadmap is built beforehand. Find a path from start to goal then finish
- Query is slower.
- No saved roadmap - better memory
- Can deal with dynamic environment

2.1 Probabilistic Road Map (PRM)

Probabilistic Roadmap (PRM) is a multi-query algorithm.
There are 2 steps:
1. Preprocessing state: build roadmap
2. Query state: search roadmap given start and goal

2.1.1 Roadmap Construction Algorithm

Algorithm 1 Roadmap_Construction_Algorithm

Input:
      n: number of nodes to put in the roadmap
      k: number of closest neighbors to examine for each configuration
Output:
      A roadmap $G = (V, E)$
 1: $V \leftarrow \varnothing$
 2: $E \leftarrow \varnothing$
 3: **while** $|V| < n$ **do**
 4:    **repeat**
 5:      $q \leftarrow$ a random configuration in Q
 6:    **until** q is collision-free
 7:    $V \leftarrow V \cup \{q\}$
 8: **end while**
 9: **for all** $q \in V$ **do**
10:    $N_q \leftarrow$ the k closest neighbors of q chosen from V according to *dist*
11:    **for all** $q' \in N_q$ **do**
12:      **if** $(q, q') \notin E$ **and** $\Delta(q, q') \neq NIL$ **then**
13:        $E \leftarrow E \cup \{(q, q')\}$
14:      **end if**
15:    **end for**
16: **end for**

Let $\Delta(q, q')$ be a function that returns either a collision-free path from q to q' or NIL if it cannot find such a path. There are many algorithms can be used for Δ. For example, if we want to check whether there is a straight-line path from q to q, we can use binary search to check for collision.
Roadmap construction algorithm shows in Algorithm 1:
- Line (1) and (2): Initialize graph $G = (V, E)$ to be empty.
- Line (3) to (8) describes sampling process: keep sampling to get n numbers of collision-free samples.

- Line (9) to the end is process to build roadmap from n samples: for every sample node $q \in V$, create a set N_q of k closest neighbors. Then from every $q' \in N_q$, whenever function $\Delta(q, q')$ succeeds to compute a collision-free path from q to q', the edge (q, q') is added to E. Figure 8 shows an example of roadmap using Δ as straight-line planner
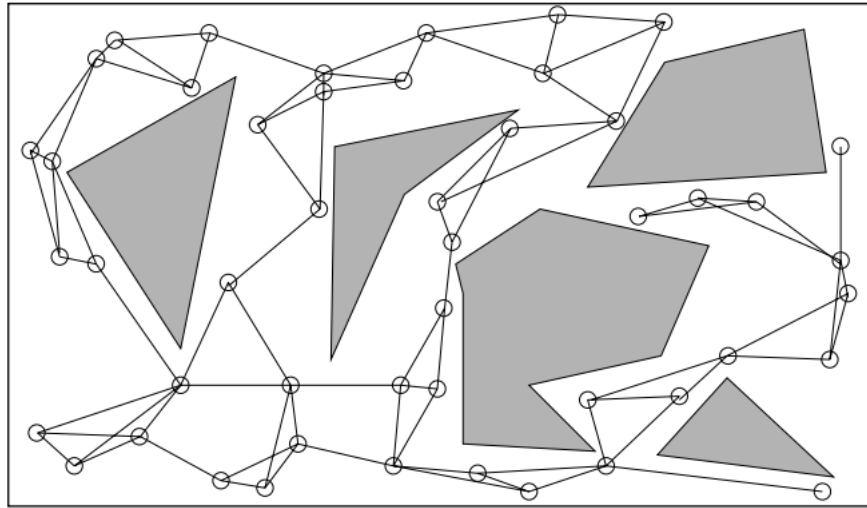

Figure 8: An example of a roadmap

2.1.2   Query

When you have a start point and end point, you connect them to the graph and then using graph searching algorithm to find path from start to end.

2.1.3   Failure

If obstacles are very closed to each other, generated roadmap will be disconnected. Possible solution is increasing sampling rate to get more samples. However, probability to sample between the obstacles is very unlikely. For example, in figure 9, the chance to sample points in the narrow road between 2 obstacles is very low. If you keep sampling more points, the graph will become so dense and will take a lot of time to find a path between start and end point.

 Solution: sample more points toward edges of the obstacles. We can use binary search to find the closest point to the obstacle that's collision-free, and then sample more points near that point. The process of sampling becomes more expensive, but result graph is less dense and faster to compute path.
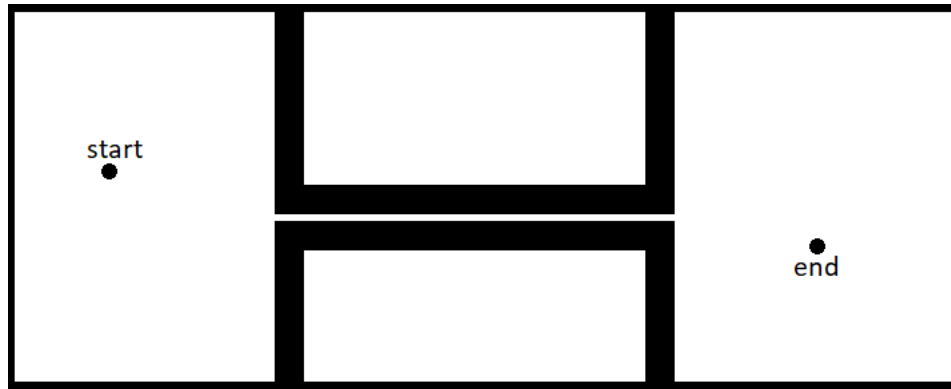
Figure 9: An example when obstacles are very closed to each other

2.2   Visibility PRM

The roadmap is constructed incrementally by randomly sampling the configuration space and attempting to connect some pairs of collision-free samples by the local method. The visibility roadmaps are build without any explicit computation of the visibility domains.
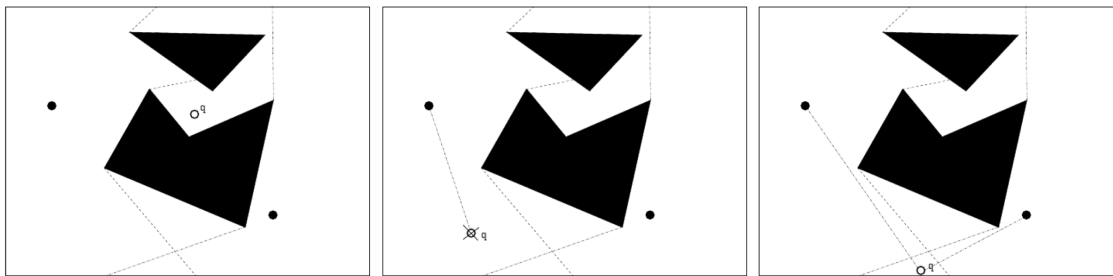

Figure 10: node added as a guard; node rejected; connection node merging two connected components

2.2.1   Principle

The algorithm that we propose below is general. It allows us to build visibility roadmaps without requiring any explicit computation of the visibility domains. The roadmap is constructed incrementally by randomly sampling the configuration space and attempting to connect some pairs of collision-free samples by the local method. Figure 10 illustrates the principle of the sampling strategy used at each iteration of the algorithm. Randomly chosen configurations are checked for collision to generate samples in C_{free}; when a free

sample is found, it is added to the roadmap either if it does not 'see' any another node of the current roadmap (i.e. it is a new guard) or if it is seen by at least two nodes belonging to two distinct connected components of the roadmap (i.e. it is a connection node). The end of the roadmap's construction is controlled by a termination condition related to the volume of free space currently covered by the roadmap.

2.2.2   Guard and connection node

When a free sample is found, it is added to the roadmap in two cases:
- If it does not "see" another node already in R . This will be a new guard.
- If it is seen at least by two nodes belonging to two distinct connected components of will be a connection node.

2.2.3   Algorithm

The algorithm 2, called Visib-PRM, iteratively processes two sets of nodes: Guard and Connection. The nodes of Guard belonging to a same connected component (i.e. connected by nodes of Connection) are gathered in subsets G_i.

At each elementary iteration, the algorithm randomly selects a collision-free configuration q. The main loop processes all the current components G_i of Guard. The algorithm loops over the nodes g in G_i , until it finds a node visible from q. The first time the algorithm succeeds in finding such a visible node g, it memorizes both g and its component G_i, and switches to the next component G_{i+1}. When q 'sees' another guard g' in another component G_j , the algorithm adds q to the Connection set and the component G_j is merged with the memorized G_i. If q is not visible from any component, it is added to the Guard set. The main loop fails to create a new node when q is visible from only one component; in that case q is rejected.

Parameter *ntry* is the number of failures before the insertion of a new guard node. $1/ntry$ gives an estimation of the volume not yet covered by visibility domains. It estimates the fraction between the non-covered volume and the total volume of C_{free}. This is a critical parameter which controls the end of the algorithm. Hence, the algorithm stops when *ntry* becomes greater than a user set value M, which means that the volume of the free space covered by visibility domains becomes probably greater than (1-1/M).

2.2.4   Failure

The random generation of the guards may produce in some cases guards that will be difficult to connect. This effect is illustrated in Fig. 11 where two guards have been generated near the boundary of the black triangular obstacle. They fully cover C_{free}, however the intersection of both visibility domains is 'unfortunately' small. The only way to complete the roadmap is to pick a connection node in the small triangle. Then

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Algorithm 2 Visibility_PRM_Algorithm

1: *Guard* ← ∅
2: *Connection* ← ∅
3: *ntry* ← 0
4: **while** *ntry* < *n* **do**
5:     *q* ← a random configuration in *Q*
6:     g_{vis} ← ∅
7:     G_{vis} ← ∅
8:     **for all** Conponents G_i of *Guard* **do**
9:         **repeat**
10:             *found* ← *FALSE*
11:             **for all** *nodesg* ∈ g_i **do**
12:                 **if** *q* belongs to *Vis(g)* **then**
13:                     *found* ← *TRUE*
14:                 **end if**
15:                 **if** g_{vis} = ∅ **then**
16:                     g_{vis} ← *g*
17:                     G_{vis} ← G_i
18:                 **else**
19:                     Add *q* to Connection;
20:                     Create edges (*q*, *g*) and (*q*,g_{vis})
21:                     Merge components G_{vis} and G_i
22:                 **end if**
23:             **end for**
24:         **until** *found=TRUE*
25:         **if** g_{vis} = ∅ **then**
26:             add*q* to *Guard*
27:             *ntry* ← 0
28:         **else**
29:             *ntry* = *ntry* + 1
30:         **end if**
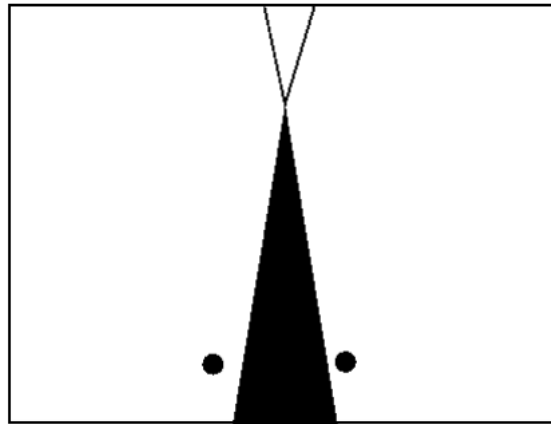31:   **end for**
32: **end while**

Figure 11: Visibility PRM Failure

the algorithm will fail if the parameter M is not sufficiently high. Nevertheless this case is only a side-effect of the algorithm. Indeed, in this example, the probability to select the first two guards with a small intersection domain is very low. Moreover, this undesirable effect was never observed in practice in all the examples we experimented on with the algorithm.