

# **PRM in Simulation**

## **Autonomous Robotics Term Project**

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### **Introduction**

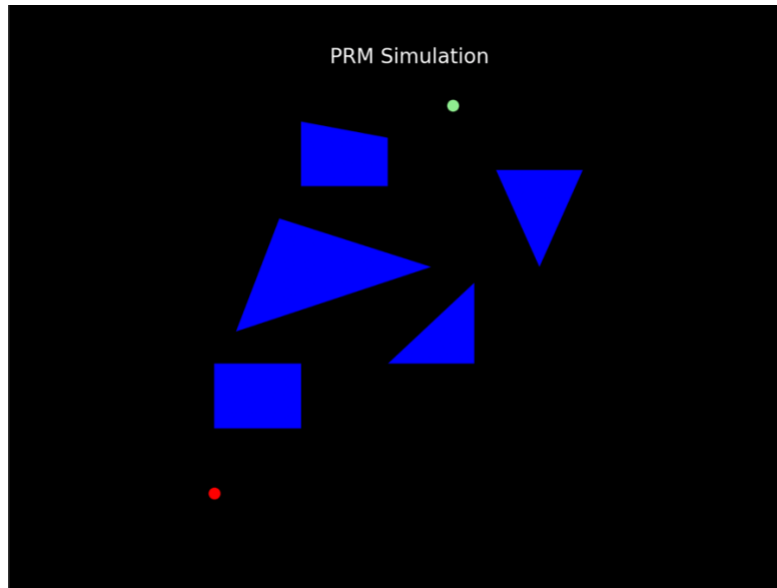
In the AI world path planning is a fundamental problem in robotics where we need to determine a safe and efficient path for a robot to navigate from a starting point to a desired destination. Probabilistic roadmap is an algorithm that is used to find optimal paths in a very complex environment. Basically, PRM uses a probabilistic approach to determine a roadmap of potential paths without contacting any obstacles to reach its goal point.

The main implementation that PRM does is by firstly randomly sampling points in the complex environment and then it uses these nodes to connect and create valid edges for the robot to travel and this connection is done in a graph like structure. The main key advantage of using PRM based path planning algorithm is that it can handle highly dynamic and cluttered environment in a very efficient way. The other main advantage is that the probabilistic nature of the algorithm allows a certain degree of flexibility and uncertainty. This makes it well suited for uncertain and changing environments.

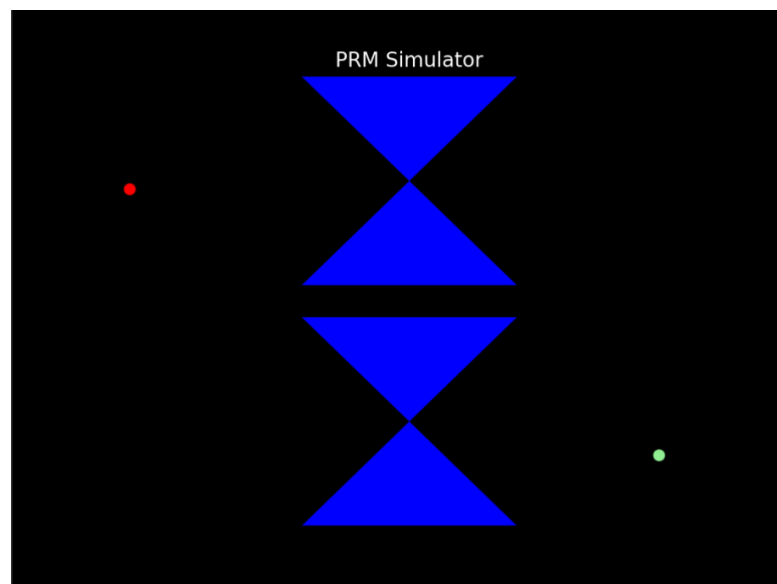
The goal of this project is trying to find the shortest path or more optimized path to reach the goal point from the start point by using the PRM implementation where we need to take random samples from the C-space and find all the edges that are not intersecting with any obstacles and finally to find a shortest path with all these edges.

### **Technical Details**

Firstly, in a particular frame I initialized obstacles of the form of Rectangles and Triangles. I took the coordinates of each of the obstacles and plotted the patch of the obstacles as you can see from the below image.



Here the red point is the start point and the green point is the goal point. We could also see that shortest distance to reach the goal point has a narrow path in-between. To perform the same task on a narrow path I've developed another obstacle map as below.



In the above there is a narrow passage from the starting Red point to the ending Green point.

Now following the PRM Pseudocode I have initialized 2 nodes  $s$  (red start node) and  $g$  (green goal node) for the roadmap. Now to sample configurations from the  $C$  space I used `random.uniform` to sample uniform coordinates. Initially I took 50 samples which includes the goal state too.

Now starting from the starting point, I would check all the sample coordinates in the  $C$ -space and find whether they intersect with any of the obstacle. I have used a simple finding intersection

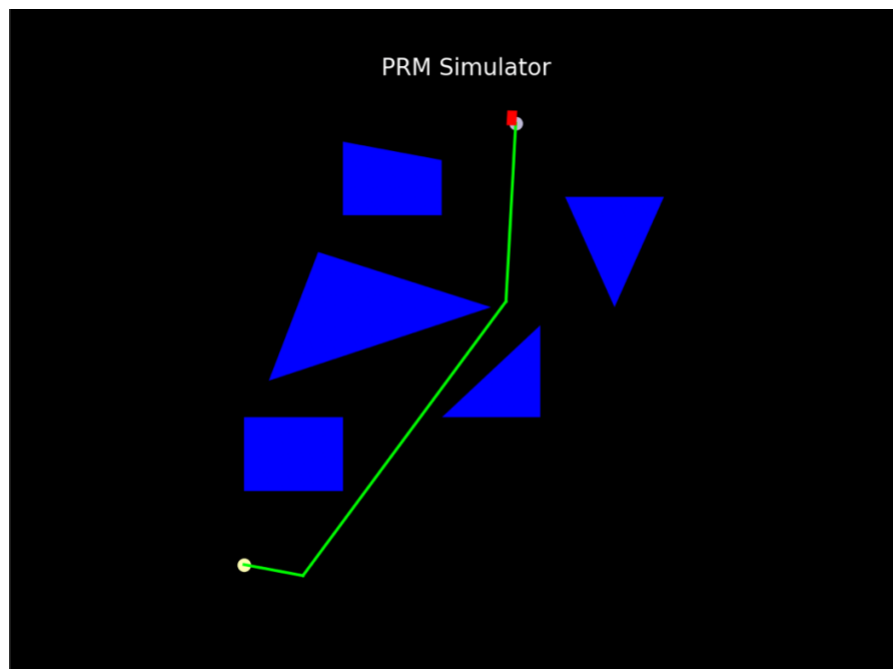
between line segments to find all the nodes. Now I keep track of all these possible edges. My dictionary structure is such a way that the keys will be all the possible nodes from the starting point and the values will be all the nodes that can be reached from that particular node. I feed I this dictionary into my path planning algorithm.

## Path planning Algorithm

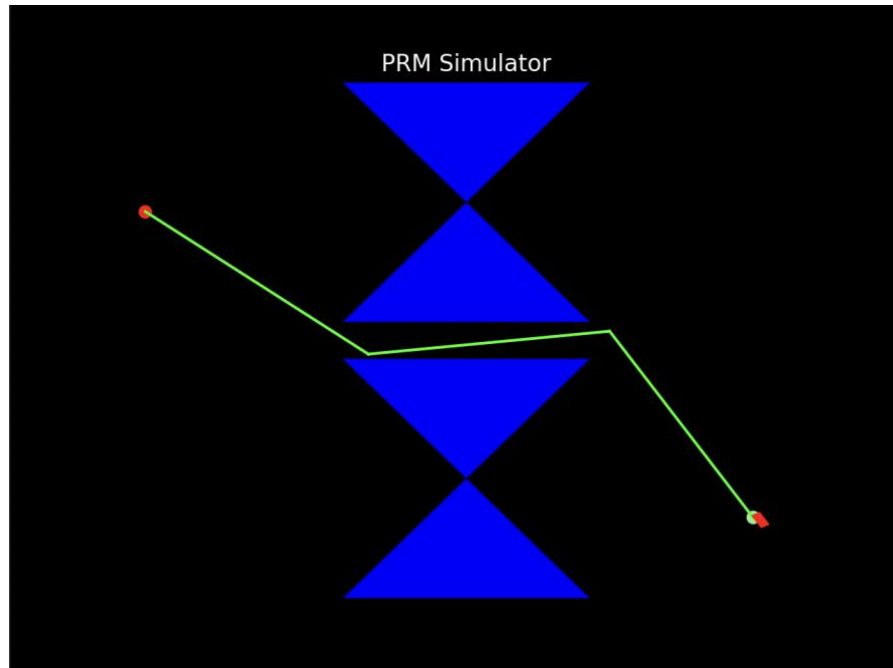
Here I tried to determine the shortest path for the robot to reach the goal state from the start point. For this part of implementation, I first tried to get all the possible nodes into a dictionary format. Then I create a heap with distance to the goal, the current node, the distance covered and the path. This way whenever I pop out a node it will be the node with the least distance to the goal point. Since I keep track of the path it has traversed, I don't need to backtrack here. Whenever the goal point is reached, I maintain the distance covered in a separate variable. This way I check for the shortest path.

## Results

Firstly, for the map with random obstacles for each sampling I get different shortest paths depending on the samples. Here I'm taking a sample of 40. Taking more than 40 samples takes a lot of time to produce the shortest path as it has to compare with all the possible nodes. But taking 40 as sample size gives the best result by finding a shortest path from the start point to the goal.



Secondly for the narrow path I provided weights when it comes to sampling. Giving higher weights to the samples near the narrow passage will give more accurate result on the shortest path. Below is the image of the RoadMap of the robot in a narrow path.



In the above RoadMap there is only a single passage towards the goal, and it is a narrow point, so by applying weights on the area around the narrow passage we could provide this result just with 20 samples.

The main issue I faced while implementing this PRM simulator is when I had to make the Rectangle head move according to the path angle. For this I calculated the velocity depending on the frame rate and the total distance and this gave me the frame rate for each edge of the shortest path. This was using  $\arctan2$  I was able to achieve the required goal by moving the rectangle in the path with appropriate angle.

Also, I believe my path planning algorithm can be optimized even more as it is taking a longer time to find the shortest path for samples more than 50. If it takes a long time, then mostly the sample couldn't find a path to the goal, and I automatically terminate the program if it takes more than 30 sec. Depending on the samples the shortest path are considered.

This course was interesting to me as I was able to gain knowledge in the area of Robotics and I always wanted to work for companies that provide solution for autonomous driving, and I believe this project have helped me a lot to grasp an idea of how autonomous driving works in real time.