Motion Planning and Trajectory Generation for Autonomous Robots

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1 Introduction

Motion planning is a fundamental aspect of autonomous robotics, enabling robots to navigate from a start position to a goal position while avoiding obstacles. This project leverages the python_motion_planning library, which provides implementations of various motion planning algorithms, to develop a complete pipeline from path planning to trajectory generation.

2 Library Overview: python_motion_planning

The python_motion_planning library, developed by ai-winter, offers a comprehensive suite of motion planning algorithms, including:

- Global Planners: Dijkstra, A*, JPS, D*, LPA*, D* Lite, (Lazy)Theta*, RRT, RRT*, RRT-Connect, Informed RRT*, Voronoi.
- Local Planners: PID, DWA, APF, LQR, MPC, RPP.
- Curve Generators: Bezier, Dubins, B-Spline, Cubic Spline, Polynomial, Reeds-Shepp.

The library is structured into modules for global planning, local planning, curve generation, and utilities, facilitating modular and extensible development of motion planning solutions.

3 Project Implementation

The project follows a structured approach:

- 1. **Environment Setup**: Define the map and obstacles.
- 2. Path Planning: Use RRT* to find a feasible path.
- 3. Path Smoothing: Apply B-Spline curves to smooth the path.
- 4. **Trajectory Generation**: Generate a time-parameterized trajectory considering kinematic constraints.
- 5. Visualization: Plot the results for analysis.

3.1 Environment Setup

An environment map of size 51x31 units is created, and obstacles are defined using rectangles and circles.

3.2 Path Planning with RRT*

The RRT* algorithm is employed to find a collision-free path from the start to the goal position.

3.3 Path Smoothing with B-Spline

The raw path obtained from RRT* is smoothed using a B-Spline curve to ensure continuity and smoothness, which is essential for trajectory generation.

3.4 Trajectory Generation

A trapezoidal velocity profile is used to generate a time-parameterized trajectory along the smoothed path, considering maximum velocity and acceleration constraints.

4 Pseudocode

The following pseudocode outlines the key components of the implementation.

4.1 RRT* Path Planning

Algorithm 1 RRT* Path Planning

- 1: **procedure** RRTSTARPLAN(start, goal, map)
- 2: Initialize tree with start node
- 3: while not reached goal do
- 4: Sample random point
- 5: Find nearest node in tree
- 6: Steer towards random point
- 7: **if** path is collision-free **then**
- 8: Add new node to tree
- 9: Rewire tree to optimize path
- 10: **return** path from start to goal

4.2 B-Spline Path Smoothing

Algorithm 2 B-Spline Path Smoothing

- 1: **procedure** BSPLINESMOOTH(path, degree, step)
- 2: Generate B-Spline curve of given degree
- 3: Sample points along the curve at specified step size
- 4: **return** smoothed path

4.3 Trajectory Generation

Algorithm 3 Trajectory Generation with Trapezoidal Velocity Profile

```
1: procedure GENERATETRAJECTORY(path, v_{max}, a_{max})
         Compute total arc length L of the path
 2:
         Compute acceleration time t_{accel} = \frac{v_{max}}{a_{max}}
 3:
         Compute distance during acceleration s_{accel} = 0.5 \cdot a_{max} \cdot t_{accel}^2
 4:
         Compute cruising distance s_{cruise} = L - 2 \cdot s_{accel}
 5:
         Compute cruising time t_{cruise} = \frac{s_{cruise}}{v_{max}}
 6:
         Compute total time T_{total} = 2 \cdot t_{accel} + t_{cruise}
 7:
         for each time t in [0, T_{total}] do
 8:
 9:
              if t < t_{accel} then
                  s(t) = 0.5 \cdot a_{max} \cdot t^2
10:
              else if t < t_{accel} + t_{cruise} then
11:
                  s(t) = s_{accel} + v_{max} \cdot (t - t_{accel})
12:
              else
13:
                  t_{decel} = t - t_{accel} - t_{cruise}
14:
                  s(t) = s_{accel} + s_{cruise} + v_{max} \cdot t_{decel} - 0.5 \cdot a_{max} \cdot t_{decel}^2
15:
              Compute position at arc length s(t)
16:
         return trajectory
17:
```

5 Results

The implementation successfully generates a smooth and feasible trajectory for the robot. The RRT* algorithm efficiently finds a path, which is then smoothed using B-Spline curves. The trajectory generation considers kinematic constraints, resulting in a realistic motion profile.

Curve and Result Figures

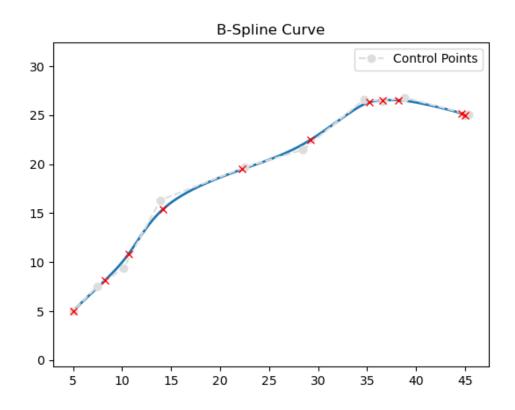


Figure 1: B-Spline Curve generated using the RRT* path.

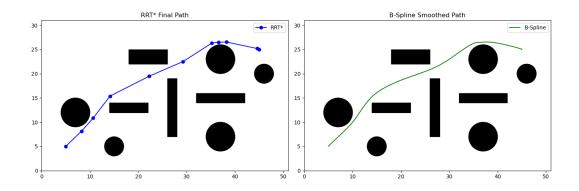


Figure 2: The left image shows RRT * path with inflated obstacles and the right image shows the B-Spline path with inflated obstacles.



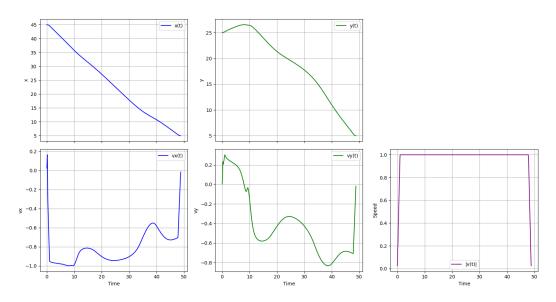


Figure 3: Motion Profile of Robot

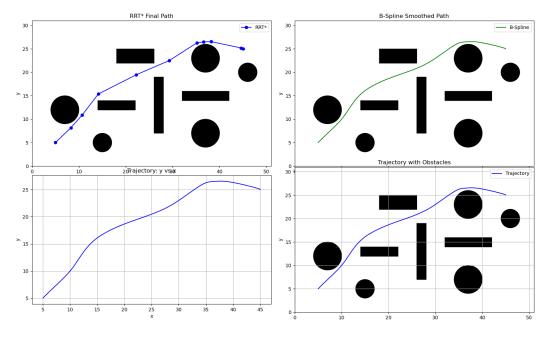


Figure 4: Bottom two images are the curve generated by plotting position vs time of robot which is derived from the motion profile of robot

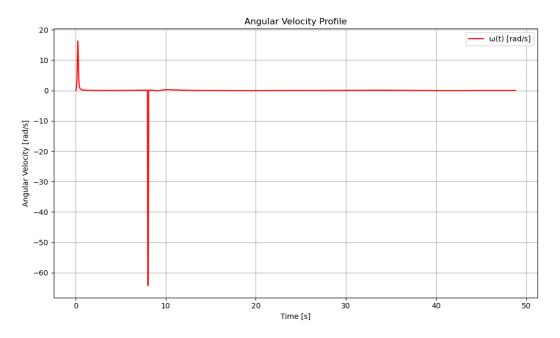


Figure 5: Angular Velocity Profile of Robot

6 Conclusion

This project demonstrates the integration of path planning, path smoothing, and trajectory generation using the python_motion_planning library. The modular approach allows for flexibility and can be extended to more complex scenarios, including dynamic environments and higher-dimensional spaces.

7 GitHub Repository

Source code of the project.

8 References

- 1. ai-winter/python_motion_planning GitHub Repository
- 2. LaValle, S. M. (2006). Planning Algorithms. Cambridge University Press.
- 3. Choset, H., Lynch, K. M., Hutchinson, S., Kantor, G., Burgard, W., Kavraki, L. E., & Thrun, S. (2005). *Principles of Robot Motion: Theory, Algorithms, and Implementations*. MIT Press.