

3D Path Planning: Pruning with Constraint Satisfaction (7A)

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November 13, 2018

1 RRT* Dubins

We can combined Dubins with RRT* to get the desired results,i.e, a path satisfying non-holonomic constraints.

So took into account these Dubins Path equations into RRT* :-

$$\begin{aligned}x &= v \cos(\theta) \\ y &= v * \sin(\theta) \\ \theta &= u\end{aligned}$$

Here

(x,y) is the position of the car, θ is the direction and $u \in [-\tan \phi, \tan \phi]$

2 RRT* Reeds-Shepp

So now we can combine Reeds-Shepp with RRT* to get the desired results,i.e, a path satisfying all non-holonomic constraints. As discussed earlier the equations for a car for non-holonomic constraints were:-

$$\begin{aligned}dx &= v * \cos(\phi) * \cos(\theta) \\ dy &= v * \cos(\phi) * \sin(\theta) \\ d\theta &= (v/L) * \sin(\phi) \\ distance(p, q) &= \sqrt{(p.x - q.x)^2 + (p.y - q.y)^2 + A * \min[(p.\theta - q.\theta)^2, (p.\theta - q.\theta + 2\pi)^2, (p.\theta - q.\theta - 2\pi)^2]}.\end{aligned}$$

Here

L is the length of the car.

$A = L * L$.

After finding the nearest node (vn) of the randomized configuration (q), we have to find the optimal input(v, ϕ) to get the new node (qn) closest to q. This is a optimization problem, defined as following:-

$$\begin{aligned}distance(qn, q) &= \sqrt{(qn.x - q.x)^2 + (qn.y - q.y)^2 + A * \min^2[(qn.\theta - q.\theta), (qn.\theta - q.\theta + 2\pi), (qn.\theta - q.\theta - 2\pi)]} \\ Where, \\ qn.x &= vn.x + v * dt * \cos(\phi) * \sin(vn.\theta) \\ qn.y &= vn.y + v * dt * \cos(\phi) * \sin(vn.\theta)\end{aligned}$$

$$qn.\theta = vn.\theta + v * dt * \sin(\phi)$$

The above distance metric used is that of Reeds-Shepp algorithm. The combination of RRT* and Reeds-Shepp gives us a smooth, constraint and optimal path in real-time. Here now every time we select a random point we will also select a θ (direction) which will also decide the path to be taken.

But while implementation we faced some issues:-

- Due to very big resolutions screen the algorithm takes a lot of time to converge to a solution and also sometimes not converges to optimal path.
- Also there are a lot of paths to enumerate from to select an optimal path, which adds to the complexity.

References

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