MT Robot Reading Group

Real-Time Trajectory Replanning for MAVs using Uniform B-splines and a 3D Circular Buffer

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Agenda

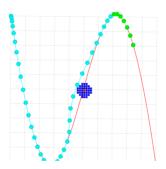
- 1. Introduction
- 2. Related Work
- 3. Approach

 - Trajectory Representation
 Mapping of the local environment
 Trajectory Optimization
- 4. Results
- 5. Discussion

Introduction

- Most system assume a static environment
- For an agent in a dynamic environment

 - Two layer approach:
 Global planner: Plan trajectory with static obstacles
 Reactive planner: Considers local information



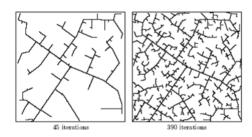
- Trajectory generation
- Map representation

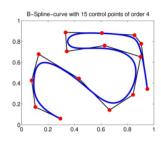
Trajectory generation:

- Search-based
- Optimization-based

Trajectory generation:

- Search-based:
 - Non-smooth path planned over a graph e.g. from RRT
 Polynomial or B-spline computed (smooth)





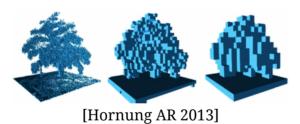
Trajectory generation:

- Optimization-based:
 - Minimize a cost function (smoothness & collision)

$$rg \min_x \lambda_s \cdot E_s(x) + \lambda_c \cdot E_c(x)$$

Map representation

- Needed to plan collision-free trajectories
 - Information about occupancy required
- Voxel grid: Simplest and most used solution



Uniform B-splines (of degree k-1)

$$p(t) = \sum_{i=0}^n p_i B_{i,k}(t)$$

where $p_i \in \mathbb{R}^n$ are control points at times $t_i, i \in [0, \dots n]$

and $B_{i,k}(t)$ are basis functions

Uniform B-splines have a fixed time interval Δt

Uniform B-splines

Advantages (compared to polynomial-splines)

• Faster optimization (fewer variables and constraints)

Disadvantages (compared to polynomial-splines)

- Trajectory does not pass through the control points For local replanning not very important

Local Environment Map

- For obstacle avoidance
- Maintain an occupancy model of the environment
- Most recent sensor measurements & some past measurements

3D Circular Buffer

• Discretize the volume into voxels



3D Circular Buffer



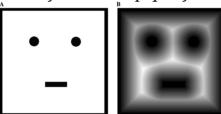
ullet Check if a voxel is in the buffer: insideVolume $(x)=0\leq x-o < N$, where o is the offset

Find address address $(n) = (n - 1) \mod N$

Local Environment Map

Distance map computation

- Compute the Euclidean distance transform (EDT)
 For fast collision checking
 Checking for collision by one look-up query



Trajectory Optimization

The local replanning problem represented as an optimization

$$E_{
m total} = E_{
m ep} + E_c + E_q + E_l$$

with

 $E_{
m ep}$ an endpoint cost function

 E_c a collision cost function

 E_q cost of the integral over the squared derivatives

 E_l soft limit on the norm of time derivatives

Trajectory Optimization

 $E_{
m ep}$ an endpoint cost function:

 $m{\cdot}$ Keep the local trajectory close to the global one $E_{
m ep}=\lambda_p(p(t_{ep})-p_{ep})^2+\lambda_v(p^{'}(t_{ep})-p_{ep}^{'})^2$

 E_c a collision cost function:

• Penalizes the trajectory point that are close to obstacles

Trajectory Optimization

 E_q cost of the integral over the squared derivatives:

• For smoothness

 E_l soft limit on the norm of time derivatives:

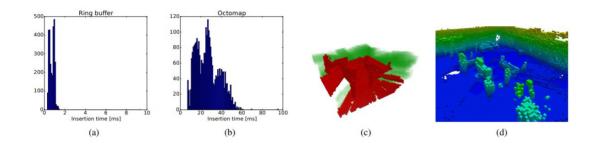
• To ensure that velocity, acceleration and higher derivatives of position remain bounded.

Implementation

- First a global trajectory is computed
- ullet In every iteration the endpoint constraints are set to the position and velocity of the global trajectory at t_{ep}
- The collision cost is evaluated using the circular buffer

Results

Circular Buffer Performance



Results

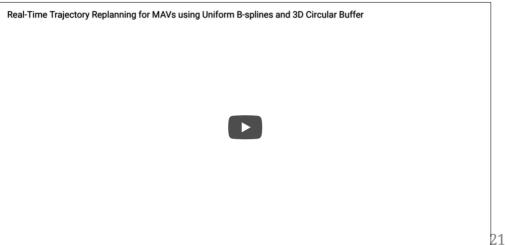
Optimization performance & System simulation

Algorithm	Success Fraction	Mean Norm. Path Length	Mean Compute time [s]
Inf. RRT* + Poly	0.9778	1.1946	2.2965
RRT Connect + Poly	0.9444	1.6043	0.5444
CHOMP $N = 10$	0.3222	1.0162	0.0032
CHOMP $N = 100$	0.5000	1.0312	0.0312
CHOMP $N = 500$	0.3333	1.0721	0.5153
[17] S = 2 jerk	0.4889	1.1079	0.0310
[17] S = 3 vel	0.4778	1.1067	0.0793
[17] S = 3 jerk	0.5000	1.0996	0.0367
[17] $S = 3$ jerk + Restart	0.6333	1.1398	0.1724
[17] $S = 3 \operatorname{snap} + \operatorname{Restart}$	0.6222	1.1230	0.1573
[17] $S = 3 \text{ snap}$	0.5000	1.0733	0.0379
[17] S = 4 jerk	0.5000	1.0917	0.0400
[17] $S = 5$ jerk	0.5000	1.0774	0.0745
Ours C = 2	0.4777	1.0668	0.0008
Ours $C = 3$	0.4777	1.0860	0.0011
Ours $C = 4$	0.4888	1.1104	0.0015
Ours $C = 5$	0.5111	1.1502	0.0021
Ours $C = 6$	0.5555	1.1866	0.0028
Ours $C = 7$	0.5222	1.2368	0.0038
Ours $C = 8$	0.4777	1.2589	0.0054
Ours C = 9	0.5777	1.3008	0.0072

Operation	Computing 3D points	Moving volume	Inserting mea- sure- ments	SDF compu- tation	Trajectory opti- mization
Time [ms]	0.265	0.025	0.518	9.913	3.424

TABLE II: Mean computation time for operations involved in trajectory replanning in the simulation experiment with depth map measurements sub-sampled to 160×120 and seven control points optimized.

Results



Something for us?

Differences:

- 2D sensor & map Car like robot (2D)

Similarities:

- Need for replanning (dynamic environment)
- Sort of local map already in use

Discussion