
Real-Time 3D Navigation for Autonomous Vision-Guided MAVs

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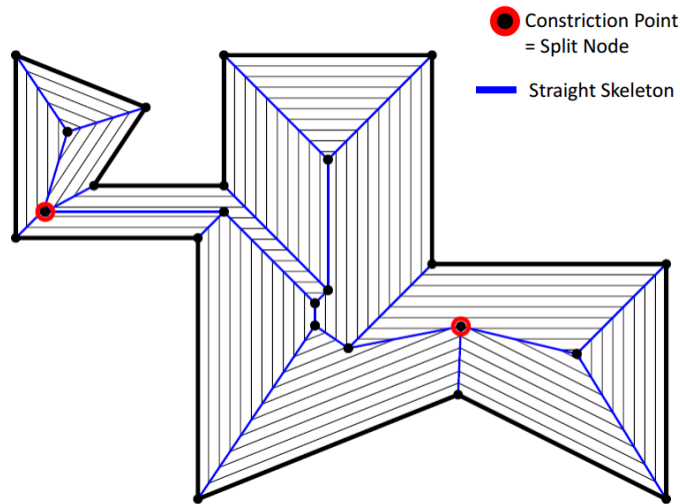
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Paper Presentation #2

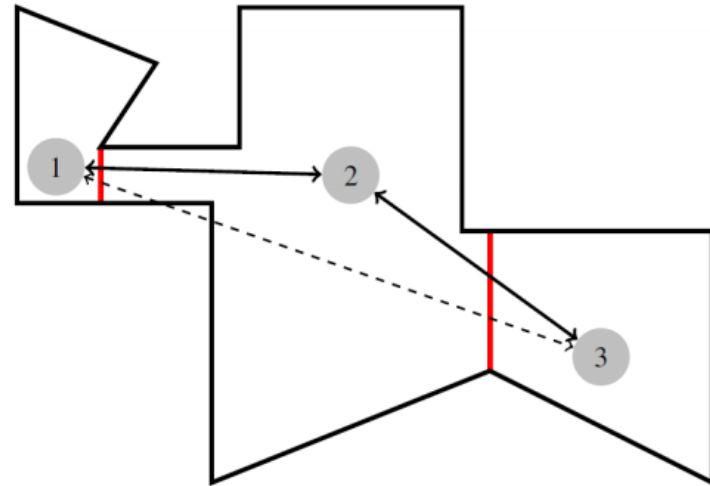
KAIST

The KAIST logo consists of the letters 'KAIST' in a bold, blue, sans-serif font. Below the text is a light blue, horizontal oval shape that serves as a shadow or base for the letters.

Suzi Kim's Presentation



Cell Decomposition
- Shrink
- Split



Cell Visit
Using TSP

Contents

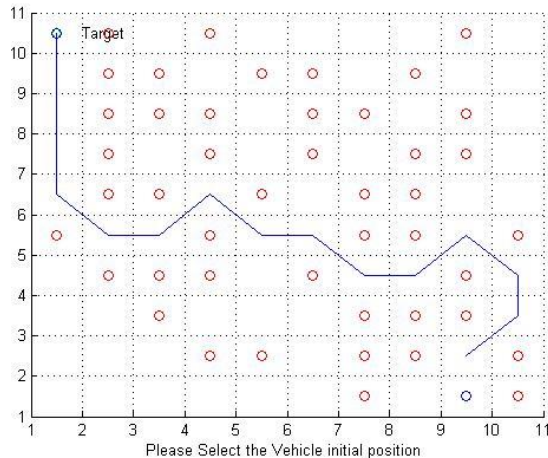
- **Introduction**
- **Conventional approach**
- **Basic concepts**
- **Detail of each concepts**
- **Result**

Introduction

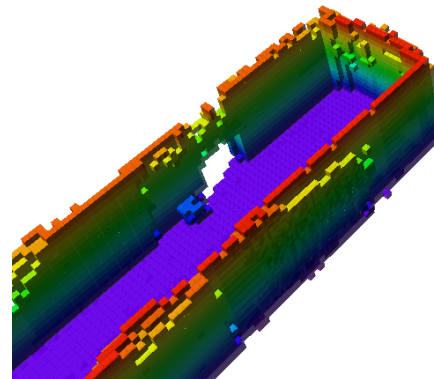
- Regular 3D state lattice requires a **large amount of memory** while graph search even though problem is **easier to solve**.
- Using Octree-based state lattice which represent discretizes **large swathes of free space into few symbolic octants**.
- **Warning!**
 - It does **not contain any Math**, just in robotical perspective!
 - So, just basic result comparison with conventional method.

Conventional Approach

- There are several grid-based path planning method in 2D.
- In 3D, there are too many points, so reduced them by using Octomap.
- Using reduced 3D grid, Researchers can use conventional A* or other algorithms



Grid based pathplanning



Octomap

Basic Concepts

- **Simplify Quadrotor dynamic**
- **Reduce resolution of Octomap (octants)**

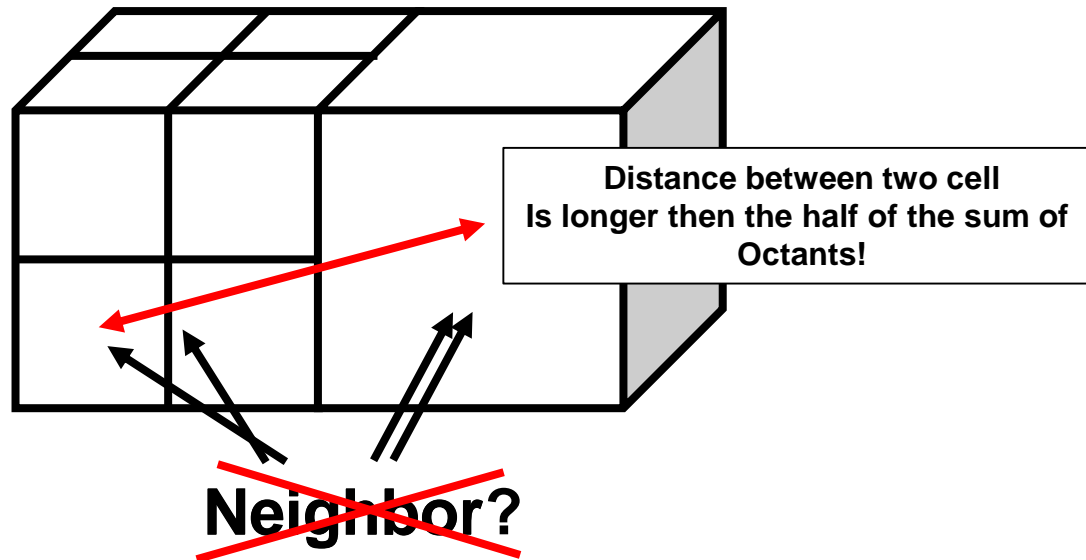
- **Octree-Based State Lattice**
 - Adjacency between octree node states
 - Multi-resolution path lookup-table
 - Pre-discretization

- **Local 3D State Lattice**

- **Graph search**
 - Optimal path finding
 - Path reconstruction

Octree-Based State Lattice

- Adjacency between octree node states
 - To determine whether two octants are adjacent to each other.
 - If distance between two cell's center exceeds half of the sum of two octants' cell size, two octants are not adjacent.



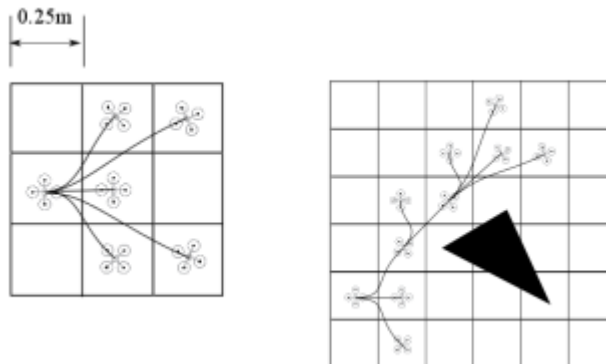
Algorithm 1 Algorithm to find neighbors for each of a *node*'s children when the node is split.

```
1: if node is no longer a leaf due to updated map information then  
2:   Split node into eight children (child[i], i ∈ 1, 2, ..., 8)  
3:   for i = 1 → 8 do  
4:     child-node = child[i]  
5:     add the child-node's brothers (child[j], j ≠ i, j ∈ 1, 2, ..., 8) as neighbors  
6:     for each neighbor of node, neighbor[k] do  
7:       if neighbor[k] is still the neighbor of child-node after the split then  
8:         add neighbor[k] as the neighbor of child-node  
9:       end if  
10:    end for  
11:  end for  
12: end if
```

Octree-Based State Lattice

- **Multi-resolution path lookup-table**

- **Computing path between every octants' consume much computational cost.**
- **How about save all pre-computed cost and path in the table?**
- **They set 16 states in yaw angle (22.5 deg inc)**
- **They set lookup index ($\theta_1, x_1-x_2, y_1-y_2, z_1-z_2, \theta_2$)**



Algorithm 2 Multi-resolution path lookup-table construction.

```

1: for  $i = 1 \rightarrow 16$ ,  $x = -N \rightarrow N$ ,  $y = -N \rightarrow N$ ,
    $z = -N \rightarrow N$ ,  $j = 1 \rightarrow 16$  do
2:   LUT_COST[ $i$ ][ $x$ ][ $y$ ][ $z$ ][ $j$ ] = infinity
3:   LUT_PATH[ $i$ ][ $x$ ][ $y$ ][ $z$ ][ $j$ ] = undefined
4: end for
5: for  $i = 1 \rightarrow 16$  do
6:   for every state  $v$  in the state lattice do
7:     dist[ $v$ ] := infinity
8:     previous[ $v$ ] := undefined
9:   end for
10:   $Q$  := empty priority queue
11:   $s\_start$  := the origin of the lattice with an orientation
     index  $i$ 
12:  dist[ $s\_start$ ] = 0
13:  insert  $s\_start$  into  $Q$ 
14:  while  $Q$  is not empty do
15:     $u$  := vertex in  $Q$  with minimum dist[ $u$ ]
16:    remove  $u$  from  $Q$ 
17:    for each neighbor  $v$  of  $u$  do
18:       $j$  := the orientation index of  $v$ 
19:       $(dx, dy, dz)$  := the 3D coordinate difference
         between  $u$  and  $v$ 
20:      checkdist := dist[ $u$ ] + cost( $u, v$ )
21:      if checkdist < dist[ $v$ ] then
22:        dist[ $v$ ] := checkdist
23:        previous[ $v$ ] :=  $u$ 
24:        LUT_COST[ $i$ ][ $dx$ ][ $dy$ ][ $dz$ ][ $j$ ] = checkdist
25:        waypoint =  $v$ 
26:        clear LUT_PATH[ $i$ ][ $dx$ ][ $dy$ ][ $dz$ ][ $j$ ]
27:        while waypoint  $\neq s\_start$  do
28:          push back waypoint to
             LUT_PATH[ $i$ ][ $dx$ ][ $dy$ ][ $dz$ ][ $j$ ]
29:          waypoint = previous[waypoint]
30:        end while
31:      end if
32:    end for
33:  end while
34: end for

```


Octree-Based State Lattice

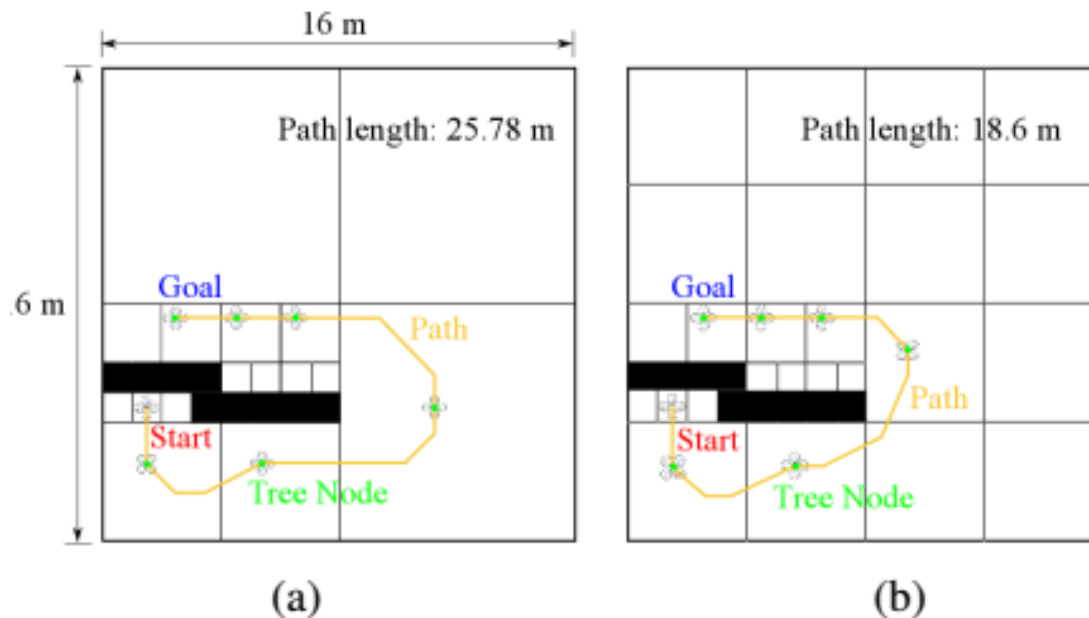
- **Multi-resolution path lookup-table**
 - **But save all computation result consume lots of memory!**
 - **They just consider 'distance' as cost.**
 - **So, $(0,0,0, \theta_1)$ to (x,y,z,θ_2) can be reflected to $(0,0,0, \theta_1)$ to $(x,y,-z, \theta_2)$!**
 - **Also, all 16 possible θ_1 can bereduced to 0,22.5,45 degrees.**
 - **So, they say memory requirement reduced by 90%**

Algorithm 2 Multi-resolution path lookup-table construction.

```
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        between  $u$  and  $v$   
20:      checkdist := dist[ $u$ ] + cost( $u,v$ )  
21:      if checkdist < dist[ $v$ ] then  
22:        dist[ $v$ ] := checkdist  
23:        previous[ $v$ ] :=  $u$   
24:        LUT_COST[ $i$ ][ $dx$ ][ $dy$ ][ $dz$ ][ $j$ ] = checkdist  
25:        waypoint =  $v$   
26:        clear LUT_PATH[ $i$ ][ $dx$ ][ $dy$ ][ $dz$ ][ $j$ ]  
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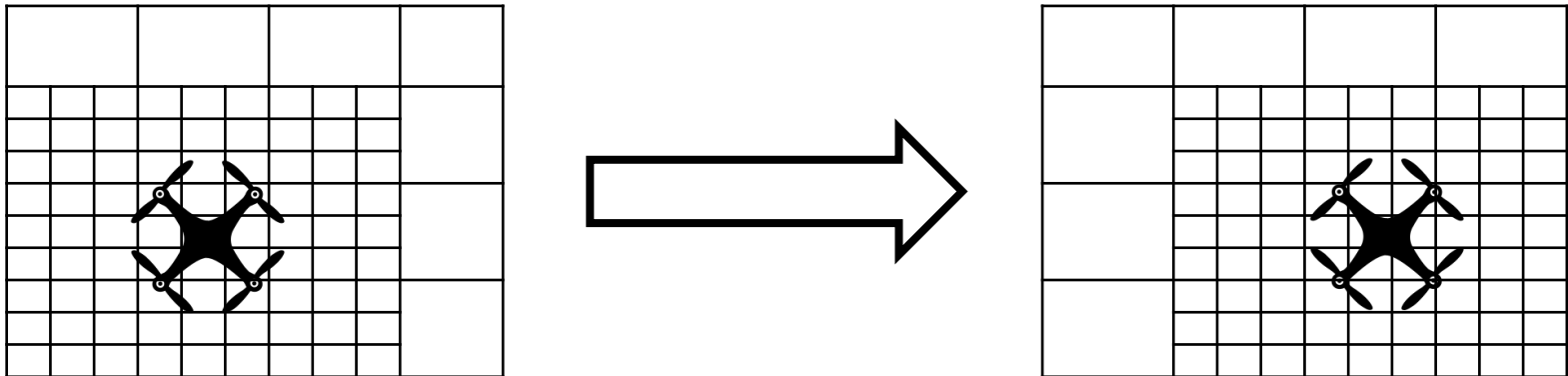
Octree-Based State Lattice

- Pre-discretization
 - Octree-based state lattice may compute **highly suboptimal path**.
 - More octree level means large pre-computed cost and path table.
 - So they enforce a minimum octree level on all leaf node.



Local 3D State Lattice

- Path planning is critical especially for obstacle avoidance.
- They make local high-resolution state lattice centered on the MAV.



- These method can maintain octree-based graph structure.
- Can help the MAV navigate around nearby obstacles.

Graph Search

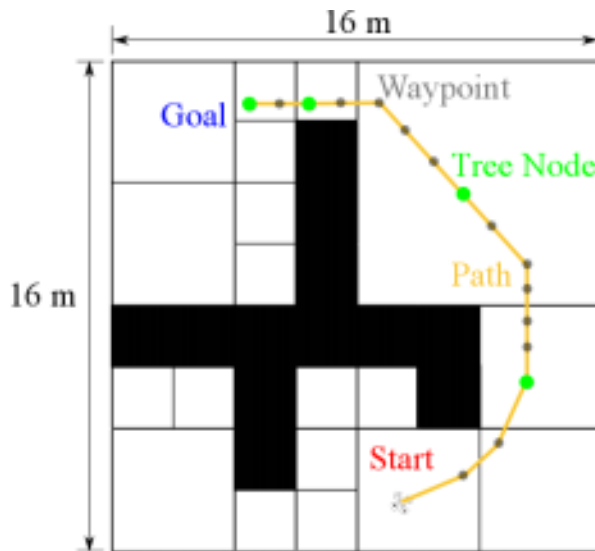
- **Optimal Path Finding**

- **Use simple A* graph search algorithm. (Using the method above, any A* based algorithm can be used)**
- **A* algorithm heavily depends on the quality of the heuristic function.**
 - **They applied holonomic-with-obstacles heuristic [1]**
 - Ignores the non-holonomic nature of robot, and then make 2D path with obstacle map
 - 3D space into 2D space by $f_2(x, y) = \min_{\theta} f_3(x, y, \theta)$. which means that 2D state is assumed to be safe (no collision) if there exists at least one safe 3D state with same 2D projection.
 - **They reduced candidate states, so A* able to find the best path in short time.**

Graph Search

- **Path reconstruction**

- **Path obtained by A* is actually a series of high-resolution primitive motion.**
- **They look up the path decompositions in the multiresolution lookup table.**



Green dot : Node achieved by A*
Grey dot : actual waypoints
Clay line : final full path

Result

- Time & Memory usage reduce
 - For 50 different goals with maximum resolution of 0.25m
 - Compare with regular-state-lattice-based path planner.

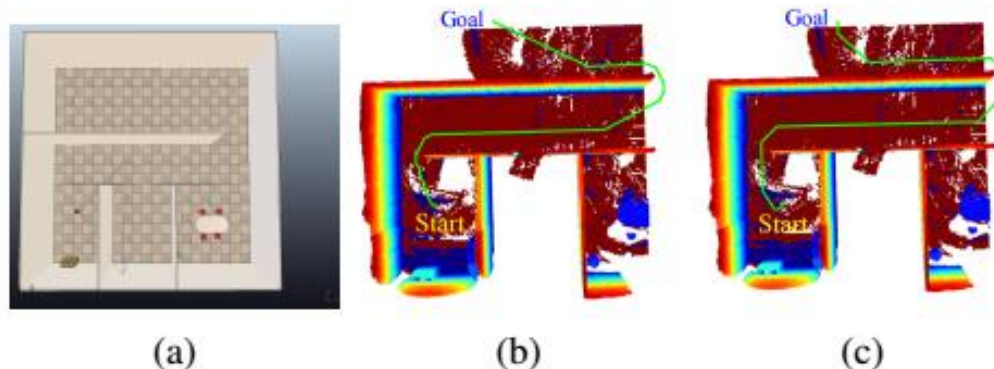
TABLE I: Statistical Results from Simulation Experiments.

	Our Path Planner	Baseline Path Planner
Map Update Time ¹ (s)	0.0991	0.0185
Graph Search Time ² (s)	0.299	10.1803
Heuristics Time ³ (s)	0.0288	0.0288
Total Time (s)	0.428	10.23
Total Path Length (m)	1108.32	1009.21
Optimality Ratio	1.11	1
Memory Usage (Gb)	0.474	1.39

¹ the time taken to update obstacle information and construct graph

² the time taken to run A* algorithm on the given graph

³ the time taken to compute heuristics



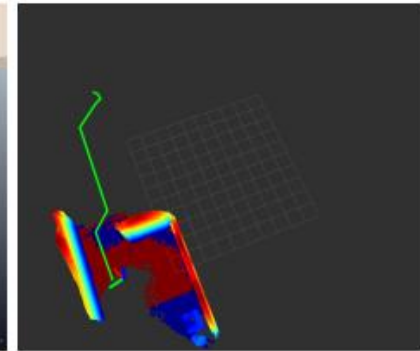
Result

- **Unknown Environment**

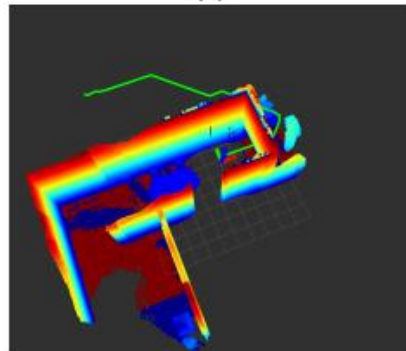
- **A : Entire Environment**
- **B : Initial Search to goal**
- **C : UAV goes through Stairs**
- **D : Successfully find path**



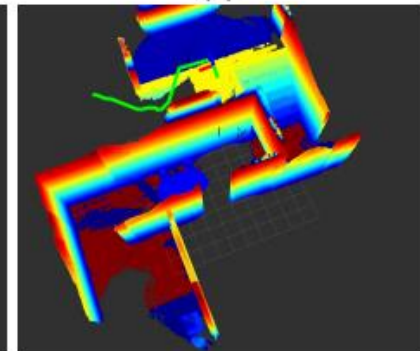
(a)



(b)



(c)

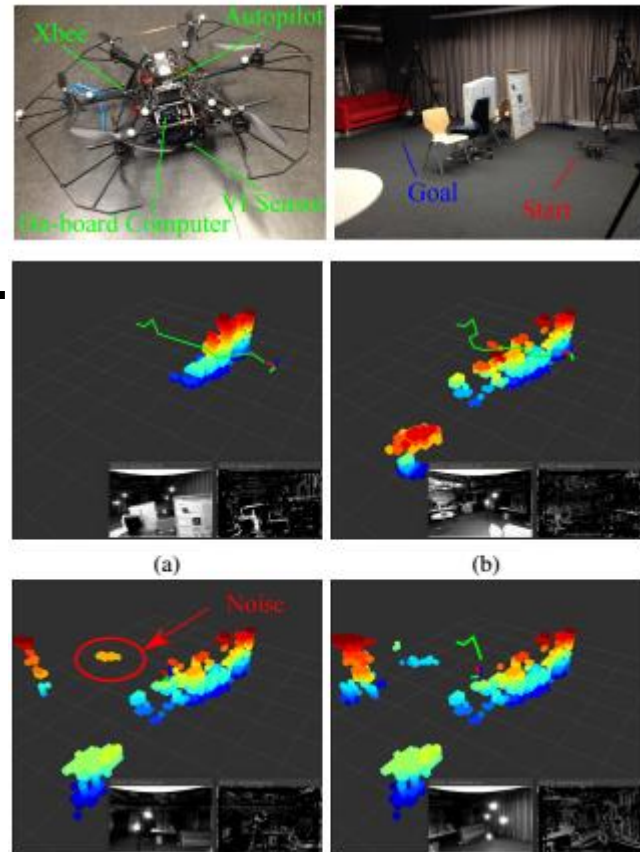


(d)

Result

- **Real Environment**

- Also in real environment, Algorithm works well.
- UAV found obstacle, and planned path.



ANY QUESTION?
