



深蓝学院
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移动机器人运动规划第二章学习分享



主讲人 JT



- 第一部分：算法
- 第二部分：仿真实现
- 第三部分：Q&A

- *Dijkstra*

$$g(n) = g(m) + Cost_{mn}$$

- A^*

$$f(n) = g(n) + h(n)$$

启发函数 h 的选择:

$$\text{Manhattan: } h(n) = \Delta x + \Delta y + \Delta z$$

$$\text{Euclidean: } h(n) = \sqrt{\Delta x^2 + \Delta y^2 + \Delta z^2}$$

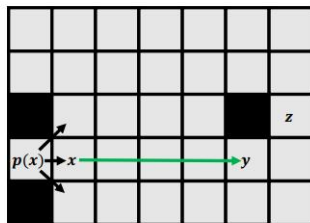
$$\text{Diagonal: } h(n) = \Delta x + \Delta y + \Delta z + (\sqrt{3} - 3) \min(\Delta x, \Delta y, \Delta z) + (\sqrt{2} - 2) [\text{mid}(\Delta x, \Delta y, \Delta z) - \min(\Delta x, \Delta y, \Delta z)]$$

● JPS

基于A*改进，一种扩展搜索节点的策略，提升复杂环境下的搜索效率。（具体见课件）

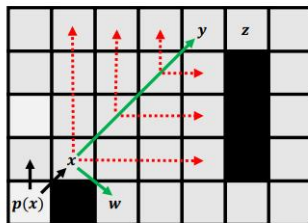
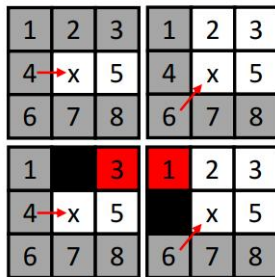


Jumping Rules



Jumping Straight

Look Ahead Rule



Jumping Diagonally

- Recursively apply **straight pruning rule** and identify y as a **jump point successor** of x. This node is interesting because it has a neighbor z that cannot be reached optimally except by a path that visits x then y.
- Recursively apply the **diagonal pruning rule** and identify y as a **jump point successor** of x.
- Before each diagonal step we first recurse straight. Only if both straight recursions fail to identify a jump point do we step diagonally again.
- Node w, a forced neighbor of x, is expanded as normal. (also push into the open list, the **priority queue**)

●A*算法在工程应用上的trick

A*算法在实际搜索的过程中会有很多代价相同的路径，这会降低搜索的效率，核心思想是打破路径间的对称性。

$$h = h + cross \times 0.001$$

$$cross = abs(\Delta x_1 \Delta y_2 - \Delta x_2 \Delta y_1 + \Delta x_1 \Delta z_2 - \Delta x_2 \Delta z_1 + \Delta y_1 \Delta z_2 - \Delta y_2 \Delta z_1)$$

$$\Delta x_1 = abs(x_n - x_g) \quad \Delta x_2 = abs(x_s - x_n)$$

$$\Delta y_1 = abs(y_n - y_g) \quad \Delta y_2 = abs(y_s - y_n)$$

$$\Delta z_1 = abs(z_n - z_g) \quad \Delta z_2 = abs(z_s - z_n)$$

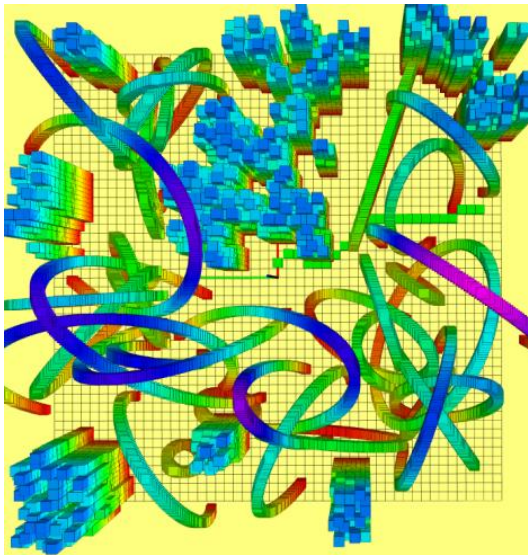
纲要

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● A* 算法框架

```
1 startNode.g = 0;
2 startNode.h = heuristicFunc(startNode, goalNode);
3 startNode.f = startNode.g + startNode.h;
4 open.insert(startNode); //起点加入open集合中
5 while(open.empty() == false){
6     currentNode = open.begin(); //从open中弹出f(n)值最小的节点n;
7     if(currentNode == goalNode){ //如果到了目标节点
8         goalNode.fatherNode = currentNode.fatherNode; //目标节点的父节点替换为当前节点, 用于回溯轨迹
9         break;
10    }
11    close.insert(currentNode); //将n加入close队列;
12    allNeighborNodes = getNeighborNodes(currentNode); //扩展n节点的邻居neighbors;
13    for(neighbor in allNeighborNodes){
14        if(neighbor not in open && neighbor not in close){ //如果邻居节点是没有访问过的节点
15            neighbor.g = currentNode.g + distance(neighbor, currentNode); //更新g值
16            neighbor.h = heuristicFunc(neighbor, goalNode); //更新h值
17            neighbor.f = neighbor.g + neighbor.h; //更新f值
18            neighbor.fatherNode = currentNode; //更新父节点
19            open.insert(neighbor);
20        }
21        else if(neighbor in open){ //如果邻居节点是在open集合中的节点
22            if(neighbor.g >= currentNode.g + distance(neighbor, currentNode)){ //如果邻居节点原本的g大于当前节点的g加上当前节点到邻居节点的距离
23                neighbor.g = currentNode.g + distance(neighbor, currentNode); //更新g值
24                neighbor.h = heuristicFunc(neighbor, goalNode); //更新h值
25                neighbor.f = neighbor.g + neighbor.h; //更新f值
26                neighbor.fatherNode = currentNode; //更新父节点
27            }
28        }
29        else if(neighbor in close){
30            continue; //对于已经在close集合中的节点不作处理, 即跳过;
31        }
32    }
33 }
34 path = backtrack(goalNode);
```

● A*运行效果

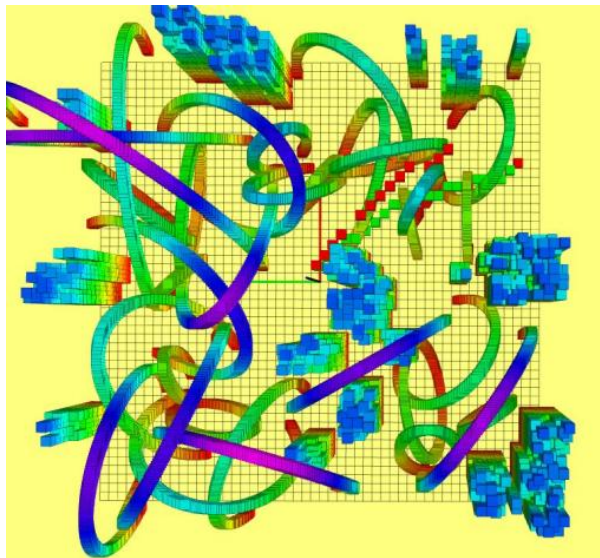


```
[ WARN] [1667206917.710253048]: 3D Goal Set
[ INFO] [1667206917.721034701]: [node] receive the planning target
[ WARN] [1667206917.722677686]: [ManhattanHeu A*]{sucess} Time in A* is 0.0986
88 ms, path cost if 1.163826 m
[ WARN] [1667206917.723258150]: visited_nodes size : 50
[ WARN] [1667206917.725409592]: [EuclideanHeu A*]{sucess} Time in A* is 1.3793
98 ms, path cost if 1.099547 m
[ WARN] [1667206917.726359106]: visited_nodes size : 844
[ WARN] [1667206917.727274119]: DiagonalHeu [A*]{sucess} Time in A* is 0.150093
ms, path cost if 1.140394 m
[ WARN] [1667206917.728217347]: visited_nodes size : 57
```

```
[ WARN] [1667218151.866045304]: 3D Goal Set
[ INFO] [1667218151.873664241]: [node] receive the planning target
[ WARN] [1667218151.875891930]: [EuclideanHeu A* without tieBreaker]{sucess} Time in A* is 1.114020 ms, path cost if 0.973697 m
[ WARN] [1667218151.877711640]: visited_nodes size : 653
[ WARN] [1667218151.879661942]: [EuclideanHeu A* with tieBreaker]{sucess} Time in A* is 1.064995 ms, path cost if 0.973697 m
[ WARN] [1667218151.880697396]: visited_nodes size : 603
```

- A*算法的运行效果如上图左边所示;
- 上图右边分别是不同启发函数和是否使用tieBreaker的运行效果。

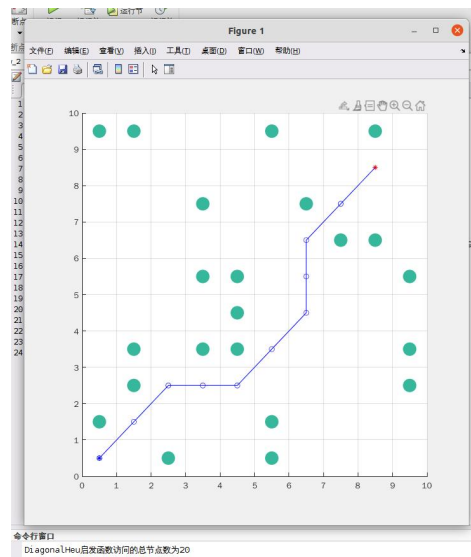
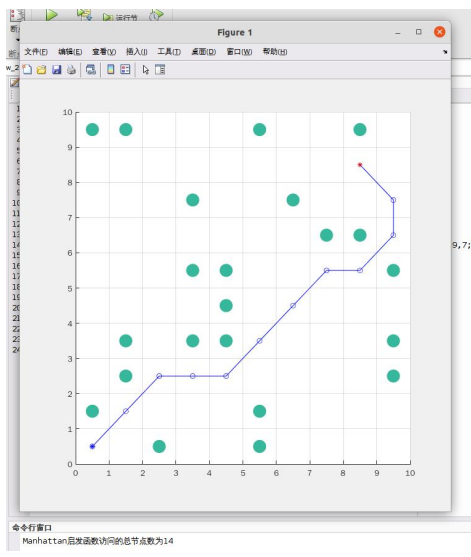
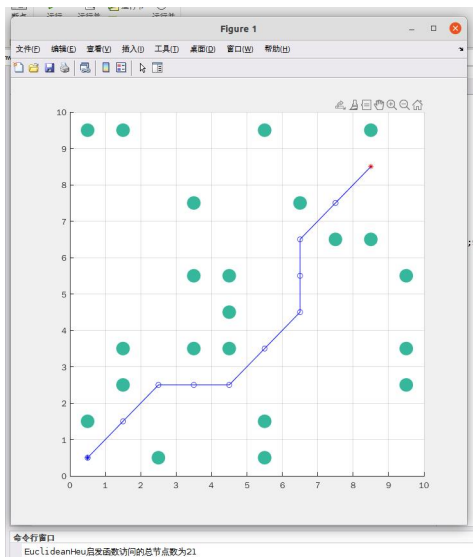
●运行效果



```
[ WARN ] [1667221241.837571875]: 3D Goal Set  
[ INFO ] [1667221241.844091374]: [node] receive the planning target  
[ WARN ] [1667221241.847724943]: [EuclideanHeu A* without tieBreaker]{sucess} Time in A* is 3.514492 ms, path cost if 1.198959 m  
[ WARN ] [1667221241.848577271]: visited_nodes size : 1063  
[ WARN ] [1667221241.855886173]: [JPS]{sucess} Time in JPS is 6.067146 ms, path cost if 4.394793 m  
[ WARN ] [1667221241.857347856]: visited_nodes size : 3966
```

- *JPS*算法的运行效果如上图左边所示;
- 上图右边是*A**算法和*JPS*运行结果对比。

●Matlab仿真结果



在matlab仿真下可以更直观地显示出不同启发式函数的路径搜索效果

● 结果分析

- 使用Manhattan距离作为启发函数得到的路径为非最优，但是访问总结点数最少，因为Manhattan函数值相对较大，更加偏向与贪心。
- 使用Euclidean距离作为启发函数得到的路径为最优，但是访问的总结点数最多，因为欧式距离为严格最小的h值。
- 使用对角距离作为启发函数得到路径为最优，且访问的总结点数相对欧式距离少。因为在栅格地图下，对角函数得到的为最小的h值，因此路径最优；对角距离值会大于欧式距离，因此访问节点数少。

- 作业内容：提交完整可编译运行的程序功能包grid_path_searcher；撰写一篇说明文档；算法流程及运行效果；对比不同启发函数对A*运行效率的影响
- 评价标准：
 - 及格：补全代码，且A*路径搜索功能正常
 - 良好：在及格的基础上对比不同启发函数对A*效率的影响
 - 优秀：在良好的基础上，加入关于JPS算法的对比，并撰写说明文档，对作业的流程进行详细的描述



感谢各位聆听 !
Thanks for Listening

