

# 移动机器人轨迹规划及运动控制

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## 实验目标

- 填补global planner节点,将所规划路径发布使其能在rviz上显示
- 结合laser信息,填补local planner节点,使其可以避障

## 实验步骤

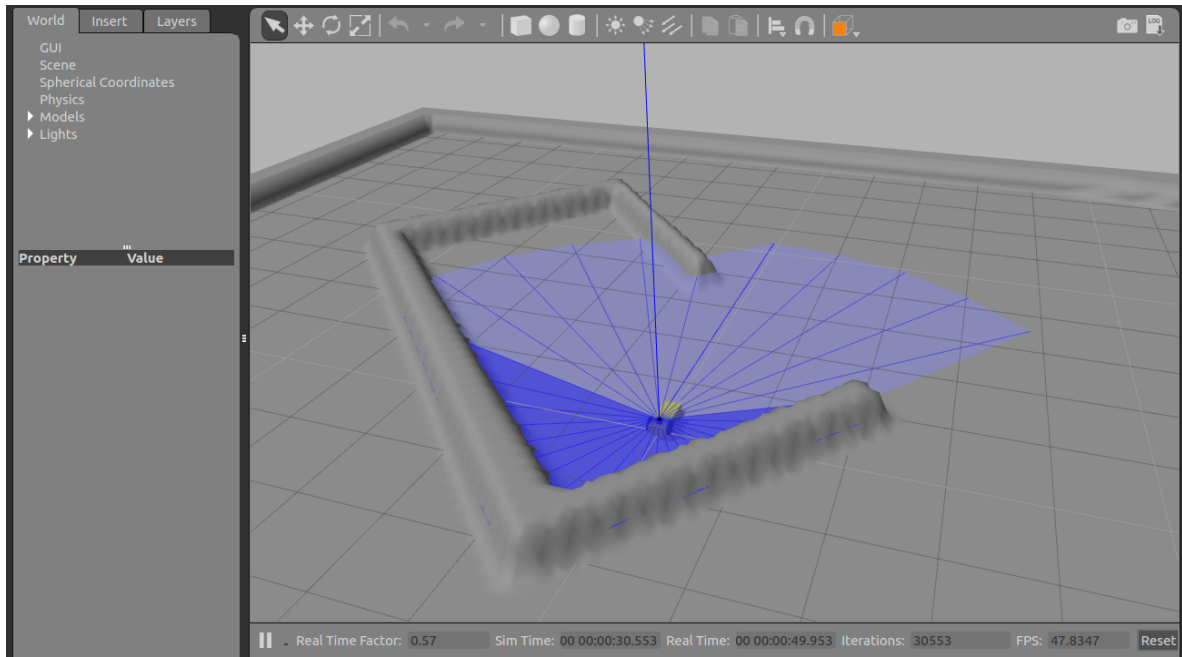
### (1) Laser 配置修改

由于电脑GPU配置不支持 `libgazebo_ros_gup_laser.so` 插件, 故使用CPU版本, 将 `urdf/course_agv.gazebo` 文件中的激光雷达配置部分修改如下:

```
1  <!-- hokuyo -->
2  <gazebo reference="course_agv__hokuyo__link">
3    <sensor type="ray" name="head_hokuyo_sensor">
4      <pose>0 0 0 0 0 0</pose>
5      <visualize>true</visualize>
6      <update_rate>3</update_rate>
7      <ray>
8        <scan>
9          <horizontal>
10             <samples>30</samples>
11             <resolution>1</resolution>
12             <min_angle>-3.14159</min_angle>
13             <max_angle>3.14159</max_angle>
14           </horizontal>
15         </scan>
16         <range>
17           <min>0.10</min>
18           <max>6.0</max>
19           <resolution>0.01</resolution>
20         </range>
21         <noise>
22           <type>gaussian</type>
23           <mean>0.0</mean>
24           <stddev>0.01</stddev>
25         </noise>
26       </ray>
27       <plugin name="gazebo_ros_head_hokuyo_controller"
28 filename="libgazebo_ros_laser.so">
29         <topicName>/course_agv/laser/scan</topicName>
30         <frameName>course_agv__hokuyo__link</frameName>
31       </plugin>
32     </sensor>
33   </gazebo>
```

将laser插件更改为了CPU版本, 并降低了激光束范围和分辨率, 否则CPU无法负载.

修改后的激光范围如图所示：



## (2) Weighted A-star

补全 `a_star.py` 文件中的部分，A\*代码结构如下：

- Maintain a **priority queue** to store all the nodes to be expanded
- The heuristic function  $h(n)$  for all nodes are pre-defined
- The priority queue is initialized with the start state  $X_s$
- Assign  $g(X_s)=0$ , and  $g(n)=\text{infinite}$  for all other nodes in the graph
- Loop
  - If the queue is empty, return FALSE; break;
  - **Remove** the node “n” with the lowest  $f(n)=g(n)+h(n)$  from the priority queue
  - Mark node “n” as **expanded**
  - If the node “n” is the goal state, return TRUE; break;
  - For all **unexpanded** neighbors “m” of node “n”
    - If  $g(m) = \text{infinite}$ 
      - Push node “m” into the queue
    - If  $g(m) > g(n) + C_{nm}$ 
      - $g(m) = g(n) + C_{nm}$
  - end
- End Loop

为了方便调整A\*中已经走过路径cost:  $g(x)$  和 heuristic-cost:  $h(x)$  权重，在 `AStarPlanner` 类中加入两个变量：

```
1 self.g_w = 1
2 self.h_w = 0.5
```

补全部分如下：

```
1 while 1:
```

```

2
3     # 1. Find the smallest in open set & Remove it
4     min_f = 1e8
5     for tinx, tnode in open_set.items():
6         if tnode.f < min_f:
7             min_f = tnode.f
8             min_inx = tinx
9
10    min_node = open_set.pop(min_inx)
11
12    # 2. Add it to closed set
13    closed_set[self.calc_grid_index(min_node)] = min_node
14
15    # 3. Is Goal?
16    if min_node.x == ngoal.x and min_node.y == ngoal.y:
17        print("Search arrive goal!")
18        ngoal = min_node
19        break
20
21    # 4. Handle all neighbor - Relaxation
22    for motion_i in self.motion:
23        tnode = self.Node(min_node.x + motion_i[0], min_node.y +
motion_i[1], 0.0, 0.0, self.calc_grid_index(min_node))
24        tnode.g = min_node.g + motion_i[2]
25        tnode.f = self.g_w * tnode.g + self.h_w * self.calc_dis(tnode,
ngoal)
26        tnode_inx = self.calc_grid_index(tnode)
27
28        # check obstacle & if in closed set
29        if self.verify_node(tnode) and not(closed_set.has_key(tnode_inx)):
30            # if not in open set -> it's a new point
31            if not open_set.has_key(tnode_inx):
32                # add it to openset
33                open_set[tnode_inx] = tnode
34
35            # if in open set
36            else:
37                if open_set[tnode_inx].g > tnode.g:
38                    open_set.pop(tnode_inx)
39                    open_set[tnode_inx] = tnode

```

### (3) DWA 动态窗口法

#### 机器人运动模型

$$\begin{aligned}
 x &= x + v\Delta t \cos(\theta_t) \\
 y &= y + v\Delta t \sin(\theta_t) \\
 \theta_t &= \theta_t + w\Delta t
 \end{aligned}$$

#### 速度采样

范围限制:

$$V_m = \{v \in [v_{\min}, v_{\max}], w \in [w_{\min}, w_{\max}]\}$$

$$V_d = \left\{ (v, \omega) \mid \begin{array}{l} v \in [v_c - \dot{v}_b \Delta t, v_c + \dot{v}_a \Delta t] \wedge \\ \omega \in [\omega_c - \dot{\omega}_b \Delta t, \omega_c + \dot{\omega}_a \Delta t] \end{array} \right\}$$

在此范围内以一定分辨率进行速度采样。

### 评价函数

$$G(v, \omega) = \sigma(\alpha \cdot \text{heading}(v, \omega) + \beta \cdot \text{dist}(v, \omega) + \gamma \cdot \text{velocity}(v, \omega))$$

#### 方位角评价：

计算达到轨迹末端时的朝向与目标之间的角度差值

$$Cost = |\theta_{traj} - \theta_{aim}|$$

```

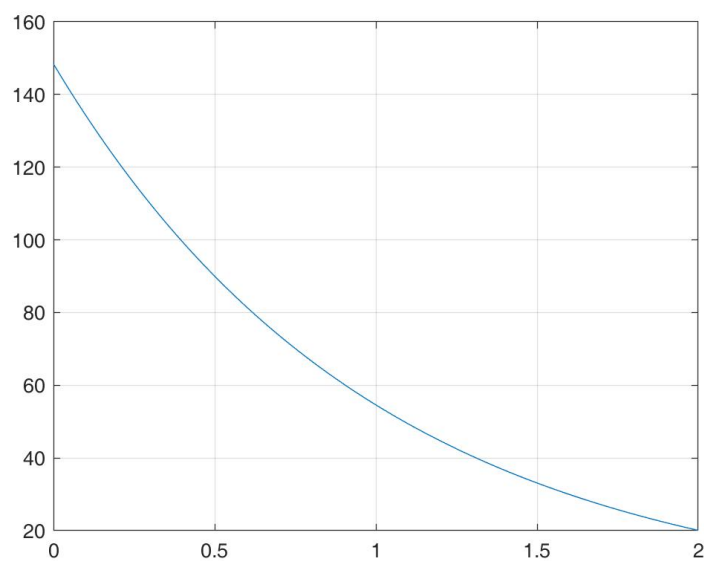
1  def calc_to_goal_cost(trajjectory, goal):
2      """
3          calc to goal cost with angle difference
4      """
5      goal_angle = math.atan2(goal[1], goal[0])
6
7      if trajectory[-1,3] >= 0:
8          cost = math.fabs(goal_angle - trajectory[-1, 2])
9      else:
10         diff = math.fabs(goal_angle - (trajectory[-1, 2]+math.pi))
11         if diff > math.pi:
12             diff = 2*math.pi - diff
13         cost = math.fabs(diff)
14
15     return cost

```

#### 障碍物距离评价：

当障碍物距离 < 1时，有如下形式的障碍物cost：

$$Cost = e^{-d+5}$$



```

1  def calc_obstacle_cost(trajjectory, ob, config):
2      """
3          calc obstacle cost inf: collision

```

```

4      """
5      if ob.size == 0:
6          cost = 0
7          return cost
8
9      ox = ob[:, 0]
10     oy = ob[:, 1]
11     dx = trajectory[:, 0] - ox[:, None]
12     dy = trajectory[:, 1] - oy[:, None]
13     r = np.hypot(dx, dy)
14     cost = 0
15     mind = r.min()
16
17     if mind < 1:
18         cost = math.exp(-(mind-1))
19
20     return cost

```

速度评价:

$$Cost = Maxv - |\bar{v}|$$

```

1  v_mean = np.mean(trajectory[:,3])
2  v_cost = 0.8 - math.fabs(v_mean)

```

**结果:**

(见 video/dwa.mp4)

从视频和命令行输出显示可以看到，小车速度已达上限，在平滑路段已以最大速度行驶，整体路径跟踪性能较好。

图中添加了2个动态障碍物，小车具有动态避障能力。

