

泛亚汽车技术中心

阶段性报告

地图模块

忻斌健

协助 丁稼毅 鞠一鸣 钱士才 李亚光

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摘要:本文讨论了地图模块的接口设计和主要的功能模块以及车辆和目标在地理坐标系下的定位....

关键词: 地图;RTK;数据融合;路径规划;微波雷达;SRR,ESR.

HD Map implementation by RTK deployment for Path Planning with Vehicle and Object localization

Xin Binjian

Abstract: This document has discussed the implementation of HD Map and the vehicle and objects localization in geographic coordinate system with RTK.

Key words: HD Map; RTK; Sensor Fusion; isomorphism; Fourier transform.

1 项目假设

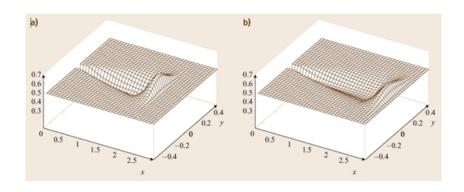
RTK, 高精度地图, 4个 SRR, 1个 ESR; 若干地标物 (landmark, beacon, 交通标示) 道路模型 (局部地图) 需要处理未知, 占据和非占据信息。接口静态的占据格栅图地图 OGM+ 动态的目标列表。根据静态地标占据格栅图用来更新本车姿态, 叠加动态的障碍物, 局部地图:障碍物列表, 车道, 静态地标, RTK 车辆定位--〉路径规划。



2 2D 世界模型

2D 占据格栅图

$$p(m|x_{1:t}, z_{1:t}) = \prod_{l=1}^{L} p(m_l|x_{1:t}, z_{1:t})$$
(1)



```
nction [veh_pose, veh_vel] = slam_ekf_patac (innerLine, middleLine, outerLine,...

l^nction [veh_pose, veh_vel] = slam_ekf_patac (innerLine, middleLine, outerLine,...

local landmarks, ...

odo_motion_x, odo_motion_y, odo_motion_yaw, ...

rtk_gps_lat, rtk_gps_lon, rtk_gps_yaw, rtk_gps_ts,...

sensor_data_raw,...

proximity)

der.extrinsic('EKF_prediction');

der.extrinsic('draw_ellipse');

g;%randn('state', 0);
```

```
10
11 determines execution and display modes
der.inline('never');
13 lobal configuration sensor;
15 rsistent sensor configuration step veh_origin_pose map ground; %step = 0;
hi2 = chi2inv(configuration.alpha, 1:1000);
17 rsistent rtk_gps_lat_last rtk_gps_lon_last rtk_gps_ts_last;
   isempty(step)
19
    % adapt to applied sensors (SRR)!
    step = 1;
22
23 nfiguration = struct('ellipses',true,'tags',false,'odometry',true, ...
                   'noise',true,'alpha',0.99,'step by step',false,...
24
                   'people', false, 'ground', 1, 'map', 2, 'observations', 3,...
25
                   'compatibility',4,'ground_hypothesis',5,'hypothesis',6,...
                   'tables',7);
27
28
    sensor.range = 5;
29
    sensor.minangle = -pi/2;
30
    sensor.maxangle = pi/2;
    sensor.srho = 0.01;
32
    sensor.stita = 0.125*pi/180;
33
   rtk_gps_lat_last =rtk_gps_lat;
35
    rtk gps lon last = rtk gps lon;
    rtk gps yaw last = rtk gps yaw;
37
    rtk\_gps\_ts\_last = rtk\_gps\_ts;
    % generate the ground data from hdmap and RTK
    ground = generate_rtk_ground(innerLine, middleLine, outerLine);
40
41
    \% start with a fresh map
42
    [map, ground] = new\_map(ground);
43
    % plot ground
45
    draw_ground(ground,landmarks, configuration);
    %%pause
47
48
```

```
% ok, here we go
50
   %%observations = get observations(ground, sensor, step);
51
   [x1,y1,utmzone,utmhemi] = wgs2utm(rtk_gps_lat,rtk_gps_lon,51,'N');
   veh origin pose.x = x1;
53
   veh\_origin\_pose.y = y1;
   veh_origin_pose.yaw = rtk_gps_yaw;
55
   veh_origin_pose.ts = rtk_gps_ts;
56
   veh_pose = veh_origin_pose;
   veh_vel = [0; 0]; \% Start point with 0 velocity.
58
   observations = get observations (ground, landmarks, veh pose, ...
59
                           sensor_data_raw, sensor, proximity);
60
   draw observations (observations, configuration, step);
61
     GT = zeros(1, size(sensor data raw,1));
63
     H = zeros(1, size(sensor data raw,1));
   map = add\_features(map, observations);
66
   % plot map
   draw_map (map, ground, configuration, step);
68
69
  \% steps = length(ground.motion);
71 Se
72
   step = step + 1;
73
   disp('-----');
74
     disp(sprintf('Step: %d', step));
76
   % EKF prediction step
77
   odometry.x = [odo_motion_x, odo_motion_y,odo_motion_yaw]'; ;%odo_motion.x;
78
   odometry.P = diag([0.25 \ 0.1 \ 5*pi/180].^2);
79
80
81
   map = EKF_prediction (map, odometry);
82
   % sense
   [x1,y1,utmzone,utmhemi] = wgs2utm(rtk_gps_lat,rtk_gps_lon,51,'N');
   veh pose.x = x1;
86
   veh pose.y = y1;
```

```
veh_pose.yaw = rtk_gps_yaw;
    veh pose.ts = rtk gps ts;
89
90
    [x2,y2,utmzone,utmhemi] = wgs2utm(rtk_gps_lat_last,rtk_gps_lon_last,51,'N');
    veh_vel = [x1-x2; y1-y2]/(rtk_gps_ts - rtk_gps_ts_last)/1000;\%ms-->s
92
93
    rtk_gps_lat_last =rtk_gps_lat;
    rtk_gps_lon_last = rtk_gps_lon;
95
    rtk\_gps\_ts\_last = rtk\_gps\_ts;
97
98
    motion.x = [x1 \ y1 \ rtk\_gps\_yaw];
    motion.P = diag([0.02 \ 0.02 \ 2*pi/180].^2); % expectation of std of RTK
100
    ground = move vehicle (ground, motion, step);
101
102
    observations = get observations(ground, landmarks, veh pose, ...
103
                             sensor_data_raw, sensor, proximity);
104
105
    % individual compatibility
    prediction = predict_observations (map, ground);
107
    compatibility = compute_compatibility (prediction, observations);
108
109
    disp(compatibility.HS);
110
    disp(compatibility.AL);
111
    \operatorname{disp}(',');
113
114
    % ground truth
115
    % your algorithm here!
116
    % 1. Try NN
    % 2. Complete SINGLES and try it
    \% 3. Include people and try SINGLES5
119
    % 4. Try JCBB
120
121
    H = NN (prediction, observations, compatibility, configuration);
123
    draw_map (map, ground, configuration, step);
124
    draw observations (observations, configuration, step);
125
126
```

```
127
    draw compatibility (prediction, observations, compatibility, configuration);
128
129
    \operatorname{disp}(',');
131
    draw_hypothesis (prediction, observations, H, 'NN:', 'b-',configuration);
132
133
    % update EKF step
134
    map = EKF_update (map, prediction, observations, H, step);
136
137
    % only new features with no neighbours
138
    new = find((H == 0) & (compatibility.AL == 0));
139
    if nnz(new)
141
       map = add features (map, observations, new);
142
    end
144
    draw_map (map, ground, configuration, step);
145
h_{pose} = map.x(1:3);
```

以下给出六种典型的数列空间和函数空间的定义,文字叙述和符号表示依照文献[1].

定义 **1** (空间 l^p ($p \ge 1$)). 一切满足 $(\sum_{i=1}^{\infty} |\xi_i|^p)^{1/p} < +\infty$ 的数列 $x = (\xi_1, \xi_2, \cdots)$ 的全体记为 l^p . 容易验证

$$\|x\|_p = (\sum_{i=1}^{\infty} |\xi_i|^p)^{1/p} < +\infty$$

是 l^p 上的范数.

. . .

3 追踪列表(动态)

```
int main() {
printf("hello, world");
return 0;
}
```

- typedef struct{
- 2 int id;
- 3 // set to 1 in the very first cycle onl that an object is output, 0 in all other cycles.
- 4 bool newObj;
- 5 //one of GCS (WGS84/UTM), LCS, CCS, VCS, SCS, ACS, ENUM tbd
- 6 int coordinate_system;
- 7 PatObjState objState;
- 8 PatObjSize objSize;
- 9 // Pedestrian/vehicle_car/vehicle_truck/unknown/... ---> ENUM tbd.
- int objClass;
- 11 //1 if the object is moving;0 if it's still;
- bool moving;
- //tracking when the object is seen by which exteroceptive sensor;
- PatTime lastSeenBySensor[NUM_SENSOR_EXTEROCEPTIVE];
- 15 //0 for invalid; 1 for valid;
- 16 float existenceProbability;
- 17 }PatObject;

18

- 19 //maximally 256 tracked objects by all sensors around the vehicle;
- 20 PatObject object_list[128];
 - 3.1 l^p 和 $L^p(\mathbf{E})$

 l^p 和 $L^p(\mathbf{E})$ 都可分....

- 4 SLAM (同时定位与生成地图)
- 4.1 函数空间与函数空间、数列空间与数列空间之间的联系

...

4.2 函数空间与数列空间的联系

引理 **1** (Riesz-Fiesher 定理). 设 $\{e_n\}$ 是 *Hilbert* 空间 **H** 中一就范正交系, $(c_1, c_2, \cdots) \in l^2$,则存在唯一的 $x \in H$ 使 $(x, e_n) = e_n$, $n = 1, 2, \cdots$ 并且 $(x, x) = \sum_{n=1}^{\infty} |c_n|^2$.

文献 [4] 给出了 $1 \le p \le 2$ 时的 $L^p(\mathbf{E})$ 上的 Fourier 变换的构造过程,并指出当p > 2 时在广义函数的意义下 $L^p(\mathbf{E})$ 仍可导入 Fourier 变换.问题在于 $p \ne 2$ 时 Fourier 变换能否构成 $L^p(\mathbf{E})$ 与 l^p 之间的保范同构.

问题 1. 完备距离空间 $S(\mathbf{E})$ 与 s, Banach 空间 $M(\mathbf{E})$ 与 m 之间是否有同构关系?更进一步,Fourier 变换及其反演公式能否推广到完备距离空间 $S(\mathbf{E})$ 与 s, Banach 空间 $M(\mathbf{E})$ 与 m

References

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