Artificial Intelligence in Control Engineering exercise

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1 Problem

2 Configuration

3 Implementation

3.1 Particle Filter

To solve the Paticle Filter problem, we implement follow below steps:

• Prediction

- Implementing a loop with M step which is numbers of particles
- Using process model and control signals u_t in **VG** which are affected by thermal noise to calculate coordinate $x_t^{[m]}$ of robot.
- Combining coordinate of robot from above step and coordinate of landmarks in \mathbf{lm} to calculate range r_t and bearing angle b_t .
- Calculating importance factor $w_t^{[m]}$ depend on probability density function fomula with μ is matrix of expected range and bearing which are from **XODO**

$$f_x(x_1, x_2, ..., x_N) = \frac{1}{\frac{N}{(2\pi)^{\frac{1}{2}}} \|\Sigma\|^{\frac{1}{2}}} exp\left(\frac{-1}{2} (x - \mu)^T \Sigma^{-1} (x - \mu)\right)$$
(1)

• Selection

- Implementing a loop with M step.
- Choosing a index in range [1, M] for $x_t^{[m]}$ with probabilities $w_t^{[m]}$.

3.1.1 Python code

```
1 #-
2 # Libraries
3 #-
4 import numpy as np
5 from scipy.io import loadmat
6 from utils.particle_filter import ParticleFilter
7
8
9 #-
10 # Parameters
11 #-
```

```
n_{particles} = 100
13 sigma_v , sigma_g = 0.5 , 3/180*np.pi
14 sigma_r , sigma_b = 0.2 , 2/180*np.pi
_{15} \text{ wb} = 4
16 time_step = 0.025
17 particle = "max"
18
19
20 #
21 #
        Main execution
22 #
23 # Load data
24 data = loadmat("data20171107.mat")
\begin{array}{lll} & \text{25 landmarks}, & X\_\text{gt}, & Z, & U = data["lm"], & data["XTRUE"], & data["Z"], & data["VG"] \\ & n\_\text{steps} = X\_\text{gt.shape}[1] \end{array}
27
29 # Create a Particle Filter instance
30 particle_filt = ParticleFilter(
         n_particles=n_particles,
31
         n_steps=n_steps,
32
33
         landmarks \! = \! landmarks \; ,
         sigma\_v{=}sigma\_v \;,
34
         sigma\_g \!\!=\!\! sigma\_g \;,
35
36
         sigma_r=sigma_r,
         sigma_b=sigma_b ,
37
        wb=wb,
38
         time_step=time_step,
39
40 )
41
42
43 # Perform loops
44 \text{ x_start} = X_gt[:, 0, \text{np.newaxis}]
{\tt 45~X\_record~,~W\_record~=~particle\_filt.loop\_over\_steps(x\_start~,~U,~Z)}
_{48} # Visualize the result
\begin{tabular}{ll} \tt 49 & mse = particle\_filt.compute\_MSE(X\_gt, X\_record, W\_record, particle) \end{tabular}
50 print ("MSE: %.6f" % (mse))
51 particle_filt.visualize(X_gt, X_record, W_record, particle)
```

3.1.2 Result

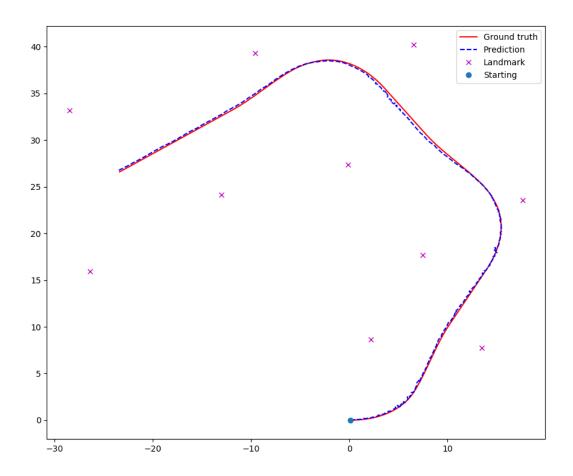


Figure 1: Trajectory