

# Robotics

## Exercise 7

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### 1 Particle Filtering the location of a car

Start from the code in `RoboticsCourse/05-car`.

The `CarSimulator` simulates a car exactly as described on slide 03:48 (using Euler integration with step size 1sec). At each time step a control signal  $u = (v, \phi)$  moves the car a bit and Gaussian noise with standard deviation  $\sigma_{\text{dynamics}} = .03$  is added to  $x, y$  and  $\theta$ . Then, in each step, the car measures the relative positions to some landmarks, resulting in an observation  $y_t \in \mathbb{R}^{m \times 2}$ ; these observations are Gaussian-noisy with standard deviation  $\sigma_{\text{observation}} = .5$ . In the current implementation the control signal  $u_t = (.1, .2)$  is fixed (roughly driving circles).

a) Odometry (dead reckoning): First write a particle filter (with  $N = 100$  particles) that ignores the observations. For this you need to use the cars system dynamics (described on 03:48) to propagate each particle, and add some noise  $\sigma_{\text{dynamics}}$  to each particle (step 3 on slide 07:23). Draw the particles (their  $x, y$  component) into the display. Expected is that the particle cloud becomes larger and larger.

b) Next implement the likelihood weights  $w_i \propto P(y_t | x_t^i) = \mathcal{N}(y_t | y(x_t^i), \sigma) \propto e^{-\frac{1}{2}(y_t - y(x_t^i))^2 / \sigma^2}$  where  $y(x_t^i)$  is the (ideal) observation the car would have if it were in the particle position  $x_t^i$ . Since  $\sum_i w_i = 1$ , normalize the weights after this computation.

c) Test the full particle filter including the likelihood weights (step 4) and resampling (step 2). Test using a larger ( $10\sigma_{\text{observation}}$ ) and smaller ( $\sigma_{\text{observation}}/10$ ) variance in the computation of the likelihood.

### 2 Gaussians

On slide 06:11 there is the definition of a multivariate ( $n$ -dim) Gaussian distribution. Proof the following using only the definition. (You may ignore terms independent of  $x$ .)

a) Proof that:

$$\mathcal{N}(x|a, A) = \mathcal{N}(a|x, A)$$

$$\mathcal{N}(x|a, A) = |F| \mathcal{N}(Fx|Fa, FAF^T)$$

$$\mathcal{N}(Fx+f|a, A) = \frac{1}{|F|} \mathcal{N}(x|F^{-1}(a-f), F^{-1}AF^{-T})$$

b) Prove:

$$\mathcal{N}(x|a, A) \mathcal{N}(x|b, B) \propto \mathcal{N}(x|(A^{-1} + B^{-1})^{-1}[A^{-1}a + B^{-1}b], (A^{-1} + B^{-1})^{-1})$$

c) Prove:

$$\int_y \mathcal{N}(x|a + Fy, A) \mathcal{N}(y|b, B) dy = \mathcal{N}(x|a + Fb, A + FBF^T)$$