Albert-Ludwigs-Universität Freiburg Lecture: Introduction to Mobile Robotics

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# Sheet 6

Topic: Particle Filter

Submission deadline: June 21, 2011 Submit to: mobilerobotics@informatik.uni-freiburg.de

#### General Notice

In the following exercises you will implement a complete particle filter. A framework containing the motion model is provided to you, so you can concentrate on the implementation of the filter itself. In this framework you have to implement the sensor model and the core components of the particle filter itself. The framework can be obtained from here. The following folders are contained in the tarball:

data This folder contains files representing the world definition and sensor readings used by the filter.

**octave** This folder contains the particle filter framework with stubs for you to complete.

plots The framework uses this folder to store images generated by the visualization.

To run the particle filter, change into the directory octave and launch the *Octave* program. Inside *Octave*, type particle\_filter to start the particle filter. Running the particle filter may take some time. While the particle filter is running, plots visualizing the state of the filter are generated and stored in the plots directory.

Note: You first have to complete all the stubs in order to get the filter working correctly.

We use the librobotics library for some of the visualization. All functions defined in the library are available in the framework.

Some implementation tips:

- Turn off the visualization to speed up the computation by commenting out the line plot\_state(... in the file particle\_filter.m.
- While debugging run the filter only for a few steps by replacing the for-loop in particle\_filter.m by something along the lines of for t = 1:50.

- The command repmat allows you to replicate a given matrix in many different ways and is magnitudes faster then using for-loops.
- When converting implementations containing for-loops into a vectorized form it often helps to draw the dimensions of the data involved on a sheet of paper.
- Many of the functions in *Octave* can handle matrices and compute values along the rows or columns of a matrix. Some useful functions that support this are sum, sqrt, and many others.

## Exercise 1: Sensor Model Implementation

Complete the file measurement model.m by implementing the sensor model of a distance-only sensor with measurement standard deviation  $\sigma = 0.2$ . Use matrix operations where possible. Instead of computing a probability it is sufficient to compute the likelihood  $p(z \mid x, l)$ .

Test your sensor model using the file test\_measurement\_model.m.

### **Exercise 2: Theoretical Considerations**

- (a) Particle filters use a set of weighted state hypotheses, which are called particles, to approximate the true state  $x_t$  of the robot at every time step t. Think of three different techniques to obtain a single state estimate  $\bar{x}_t$  given a set of N weighted samples  $S_t = \{\langle x_t^{[i]}, w_t^{[i]} \rangle \mid i = 1, \ldots, N\}$ .
- (b) How does the computational cost of the particle filter scale with the number of particles and the number of dimensions in the state vector of the particles? Why can a large dimensionality be a problem for particle filters in practice?

## Exercise 3: Resampling

A particle filter consists of three steps listed in the following:

- (a) Sample new particle poses using the motion model.
- (b) Compute weights for the new particles using the sensor model.
- (c) Compute the new belief by sampling particles proportional to their weight with replacement.

The motion model (a) is already implemented in the provided framework. In exercise 1, you implemented the measurement model (b).

To complete the particle filter framework you will have to complete the file resample.m by implementing stochastic universal sampling.

## Exercise 4: Visualization

In Exercise 2 (a) you described three ways to obtain a single state estimate from the particle cloud. Implement one of these methods in the file mean\_position.m, which is used to draw the pose of the robot in the visualization.

Note: You can use the function average\_angle.

You can generate an animation from the saved images using ffmpeg or mencoder. With ffmpeg you can use the following command to generate the animation from inside the plots folder:

ffmpeg -r 10 -b 500000 -i pf\_%03d.png pf.mp4