

## Sheet 3

Topic: Data Explanation, Sensor Models, Discrete Bayes Filter

Submission deadline: Tue 27.5.2008, 11:00 a.m. (before class)

### Exercise 1:

Consider the following situation: with a 50% probability we generate a data point from a normal distribution:

$$g(x, \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right)$$

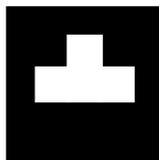
with mean  $\mu = 0$  and  $\sigma = 1$ , while with the other 50% probability we generate a data point from a normal distribution with variance 1 and adjustable mean  $\mu$ . We have observed some data points, and we wish to adjust  $\mu$  to maximize their likelihood.

Suppose further that our data points are  $a = 2$  and  $b = -1.5$ . We are told that either  $a$  or  $b$  comes from the Gaussian centered at 0 (but not both). In that case, there are only two data association possibilities: either  $a$  came from center 0 and  $b$  came from center  $\mu$ , or vice versa.

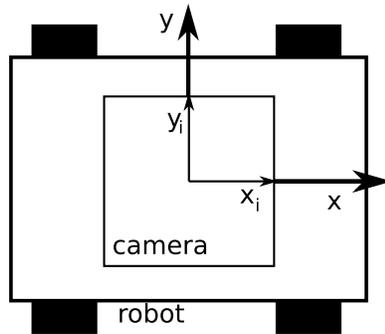
- Use gnuplot (or any other plotting tool) to plot the probability  $P(a, b | \mu)$ .
- Which is the mean  $\mu$  so that  $a$  and  $b$  have the highest likelihood?
- What is the most likely data association given the mean  $\mu$  from exercise (b). (Consider only the two data associations mentioned above.)

### Exercise 2:

Many early robots navigating using features used artificial landmarks in the environment that were easy to recognize. A good place to mount such markers is a ceiling (why?). A classical example is a visual marker: Suppose we attach the following marker to the ceiling:

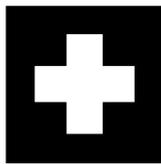


Let the world coordinates of the marker be  $(x_m, y_m, \theta_m)$ . We will denote the robot's pose by  $(x_r, y_r, \theta_r)$ . Now assume, we have a routine that can detect the marker in the image plane of a perspective camera. Let  $x_i$  and  $y_i$  denote the coordinates of the marker in the image plane, and  $\theta_i$  its angular orientation. From the setup of our system (focal length of the camera, height of the ceiling, etc.) we know that each displacement  $d$  in  $x$ - $y$ -space gets projected to a proportional displacement of  $\alpha \cdot d$  in the image plane. The relation between the camera-centric coordinate system and the robot-centric coordinate system is as depicted:



Remember that the robot looks in the direction of the positive  $x$ -axis (of the global coordinate system) if its orientation is  $\theta_r = 0$ .

- (a) Describe mathematically where to expect the marker (in global coordinates  $(x_m, y_m, \theta_m)$ ) when its image coordinates are  $(x_i, y_i, \theta_i)$  and the robot is at position  $(x_r, y_r, \theta_r)$ .
- (b) Now suppose we use several of the following symmetrical markers instead:



- (i) How many markers do we need at least in order to determine the position unambiguously?
- (ii) Sketch an unambiguous and an ambiguous placement of two such markers. How many hypotheses are there in the ambiguous case?
- (iii) Sketch an unambiguous and an ambiguous placement of three such markers. How many hypotheses are there in the ambiguous case?

### Exercise 3:

**Programming task:** A robot is living in a 1-dimensional world. It can move to the left and to the right; the world is 100m long ( $x \in [0, 100]$ ). If the robot moves, it moves approximately 4m (with Gaussian error of  $0.25m$ ). At the end of the world ( $x = 100m$ ) there is a landmark. The robot can measure the distance to the landmark with a Gaussian error of  $\sigma = 2m$ . The landmark is only visible, if the robot is close to it ( $x \geq 70m$ ).

Estimate the pose of the robot using a 1-dimensional grid. Thus, complete the method *update* in the *Grid* class. Use the following motion and sensor models:

- **Motion Model**  $p(x' | x, u)$ : Let assume that  $u \in \{+4, -4\}$ . Then, use the following motion model:

$$p(x' | x, u) = \begin{cases} 0.3750 & \text{if } x' - (x + u) = 0 \\ 0.2500 & \text{if } |x' - (x + u)| = 1 \\ 0.0625 & \text{if } |x' - (x + u)| = 2 \\ 0.0 & \text{else} \end{cases}$$

- **Sensor Model:** Use a Gaussian sensor model with standard deviation  $\sigma = 2m$ . (Thus, use only the “measurement noise” component of the “Beam-based Proximity Model” presented in the lecture.)

Remark: Make sure that the sum over all grid cells is always 1.