

Machine Learning for Robotics and Computer Vision
Winter term 2015

Homework Assignment 1

Topic 1: Introduction to Probabilistic Reasoning and Learning
October 26, 2015

Exercise 1:

In class, we had an example of a robot that can measure its distance to a wall in front of it. We modeled this using a continuous random variable with a Normal distribution $\mathcal{N}(x; \mu, \sigma^2)$.

- a) Our robot also has a camera that is not color-calibrated correctly so the color mapping is probabilistic and looks like the following table:

| Sensed color | Actual color | p(Sensed Actual) |
|--------------|--------------|--------------------|
| red | red | 0.8 |
| red | green | 0.1 |
| red | blue | 0.1 |
| green | red | 0.1 |
| green | green | 0.6 |
| green | blue | 0.2 |
| blue | red | 0.1 |
| blue | green | 0.3 |
| blue | blue | 0.7 |

Assume our robot is located in a room with 5 boxes: 2 red, 2 green and a blue one. The robot moves towards a box and it reads green. How likely is it that the box is actually green?

- b) The robot's distance sensor can be modeled using a continuous random variable with a Normal distribution with $\sigma_1 = 0.3$ m. Write the sensor model $p(z|x)$ in the full form (not the shorthand notation).
- c) Now the robot moves into another room that is empty. Initially it knows it is located at the door ($x=0$). The robot can execute *move* commands but the result of the action is not always perfect. Assume that the robot moves with constant speed v . The motion can also be modeled with a Gaussian with deviation $\sigma_2 = 0.1$ m. Write the motion model $p(x_t|x_{t-1}, u_t)$.

d) We let the robot run in the room with a speed of 1 m/s. The robot only runs forward and it updates its belief every second. Assume we obtain the following sensor measurements: ($z_1 = 1.2, z_2 = 1.6, z_3 = 2.5$).

Further assume that the position can only take discrete values from 0 to 5. Where does the robot believe it is located with respect to the door after 3 seconds? How certain is it about its location?

Exercise 2:

Try to find (for example by internet search or from the book) at least 5 examples for learning techniques that have not been discussed in class. Describe these techniques briefly and classify them with respect to the hierarchy from the lecture.

Topic 2: Regression

Exercise 3:

We are testing a tracking program. We evaluate it with the help of a quadcopter. The quadcopter sends estimates of its velocity and the tracking program estimates its global position with respect to the quadcopter's initial position (before flying).

- a) The tracker yields these tracked position estimates at a frequency of $1Hz$:

$$\mathcal{T} = \left\{ \begin{pmatrix} 2 \\ 0 \\ 1 \end{pmatrix} \begin{pmatrix} 1.08 \\ 1.68 \\ 2.38 \end{pmatrix} \begin{pmatrix} -0.83 \\ 1.82 \\ 2.49 \end{pmatrix} \begin{pmatrix} -1.97 \\ 0.28 \\ 2.15 \end{pmatrix} \begin{pmatrix} -1.31 \\ -1.51 \\ 2.59 \end{pmatrix} \begin{pmatrix} 0.57 \\ -1.91 \\ 4.32 \end{pmatrix} \right\} \quad (1)$$

Plot these data with your tool of choice (e.g. Matlab).

- b) Assuming the quadcopter flies with constant speed, which speed does it have? What is the residual error of the estimation?
- c) Now assume that the quadcopter flies with constant acceleration. What is the residual error now? Is the error higher or lower? Why?
- d) According to our last model, what is the quadcopter's most likely position in the next second?

Hint for b) and c): Use the Polynomial Regression method introduced on slides 8 - 12 of the lecture.

Exercise 4: Programming

Solve exercise 3 in your preferred programming language.

The next exercise class will take place on **November 20th, 2015**.

For downloads of slides and of homework assignments and for further information on the course see

<http://vision.in.tum.de>
