

## Sheet 4

Topic: Unscented Kalman Filter SLAM

Submission deadline: December, 3

Submit to: `robotmappingtutors@informatik.uni-freiburg.de`

### Exercise: UKF SLAM

Implement an unscented Kalman filter SLAM (UKF SLAM) system. You should complete the following three parts:

- Implement the function in `compute_sigma_points.m`, which samples sigma points given a mean vector and covariance matrix according to the unscented transform.
- Implement the prediction step of the filter by completing the function in `prediction_step.m` to compute the mean and covariance after incorporating the odometry motion command.
- Implement the correction step in `correction_step.m` to update the belief after each sensor measurement according to the UKF SLAM algorithm.

To support this task, we provide a small *Octave* framework (see course website). The above-mentioned tasks should be implemented inside the framework in the directory `octave` by completing the stubs. After implementing the missing parts, you can test your solution by running the script in `ukf_slam.m`. The program will produce plots of the robot pose and map estimates and save them in the `plots` directory.

Note that, as opposed to the EKF SLAM system you implemented in sheet 2, here the state vector and covariance matrix are incrementally grown with each newly-observed landmark. The mean estimates of the landmark poses are stacked in  $\mu_t$  in the order by which they were observed by the robot, as described in the `map` vector in the framework.

Some implementation tips:

- Be careful when averaging angles. One way to average a set of angles  $\{\theta_1, \dots, \theta_N\}$  given their weights  $\{w_1, \dots, w_N\}$  is to first compute the weighted sum of the unit-vectors of each rotation

$$\bar{x} = \sum_{i=1}^N w_i \cos(\theta_i),$$
$$\bar{y} = \sum_{i=1}^N w_i \sin(\theta_i).$$

Then the average angle  $\bar{\theta}$  is given by

$$\bar{\theta} = \text{atan2}(\bar{y}, \bar{x}).$$

- Use the following weights when recovering the mean and covariance from sampled sigma points:

$$w_m^{[0]} = w_c^{[0]} = \frac{\lambda}{n + \lambda},$$
$$w_m^{[i]} = w_c^{[i]} = \frac{1}{2(n + \lambda)},$$

where  $i = 1, \dots, 2n + 1$  and  $n$  is the dimensionality of the system.

- Turn off the visualization to speed up the computation by commenting out the line `plot_state(...` in the file `ukf_slam.m`.
- While debugging, run the filter only for a few steps by replacing the for-loop in `ukf_slam.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 than 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`, `sin`, `cos`, and many others.