

## Extended Kalman Filter for Tracking a Two-wheeled Robot

An Extended Kalman Filter is to be designed for tracking the position and orientation of a two-wheeled robot that is moving on a plane. A schematic drawing of the robot is shown in Fig. 1.

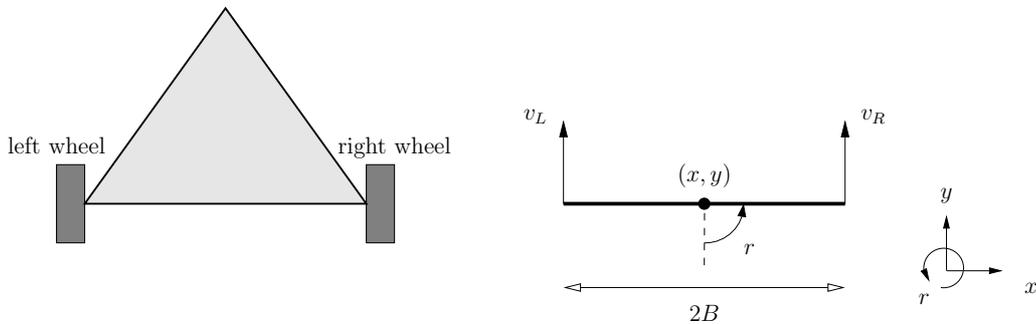


Figure 1: Top view of the two-wheeled robot (left) and relevant physical quantities (right).

The robot can command its left and right wheel angular velocities,  $u_L(t)$  and  $u_R(t)$  (in rad/s), respectively, which are assumed to be followed instantaneously. The left and right wheel radii,  $W_L$  and  $W_R$  (in m), are not known perfectly; they are modeled as random variables according to

$$\begin{aligned} W_L &= W_0(1 + \xi_L) \\ W_R &= W_0(1 + \xi_R), \end{aligned}$$

with the known nominal wheel radius  $W_0$  (in m) and the uniformly distributed random variables  $\xi_L, \xi_R \in [-\bar{\xi}, \bar{\xi}]$ . The wheel radii are assumed constant with time.

The translational speed  $v_t(t)$  (in m/s) of the vehicle is

$$v_t(t) = \frac{v_R(t) + v_L(t)}{2} \quad \text{with} \quad v_R(t) = W_R u_R(t) \quad \text{and} \quad v_L(t) = W_L u_L(t).$$

The rotational speed  $v_r(t)$  (in rad/s) of the vehicle is

$$v_r(t) = \frac{v_R(t) - v_L(t)}{2B},$$

where  $B$  is the known wheel base (distance of the wheels from the robot center), see Fig. 1. With these quantities, the kinematic equations read as follows:

$$\dot{x}(t) = v_t(t) \cos(r(t)) \tag{1}$$

$$\dot{y}(t) = v_t(t) \sin(r(t)) \tag{2}$$

$$\dot{r}(t) = v_r(t), \tag{3}$$

where  $(x(t), y(t))$  is the position of the robot (in m) and  $r(t)$  its orientation (in rad). The robot is assumed to start at  $(x(0), y(0)) = (x_0, y_0)$  with orientation  $r(0) = r_0$ , where  $x_0, y_0 \in [-\bar{p}, \bar{p}]$  and  $r_0 \in [-\bar{r}, \bar{r}]$  are uniformly distributed random variables.

At varying instances of time, the robot may receive measurements of its position and orientation that are corrupted by sensor noise, i.e.

$$\begin{aligned} z_x &= x + w_x \\ z_y &= y + w_y \\ z_r &= r + w_r, \end{aligned}$$

with  $w_x, w_y \in [-\bar{w}_p, \bar{w}_p]$ ,  $w_r \in [-\bar{w}_r, \bar{w}_r]$  uniformly distributed. At any instance of time  $t_k$ , measurements may be available from one, two, three, or none of the sensors.

All random variables  $\xi_L, \xi_R, r_0, x_0, y_0, w_x, w_y$ , and  $w_r$  are assumed to be mutually independent and independent over time.

## Objective

The objective is to design an Extended Kalman Filter to estimate the position and orientation of the two-wheeled robot. The estimator will be implemented in discrete time. At time  $t_k$ , the estimator has access to the time  $t_k$ , the control inputs  $u_L(t_k)$  and  $u_R(t_k)$ , and possibly the measurements  $z_x(t_k)$ ,  $z_y(t_k)$ , or  $z_r(t_k)$ . Furthermore, the values of all physical constants  $W_0, \bar{\xi}, B, \bar{w}_p, \bar{w}_r, \bar{p}$ , and  $\bar{r}$  are known to the estimator. The orientation, the position, and the wheel radii are estimator states.

## Provided Matlab Files

A set of Matlab files is provided on the class website. Please use them for solving the above problem.

<code>script.m</code>	Matlab script that is used to simulate the truth system, run the estimator, and display the results. <sup>1</sup>
<code>Estimator.m</code>	Matlab function template to be used for your implementation of the Extended Kalman Filter.
<code>PhysicalConstants.m</code>	Physical constants, known to the estimator.
<code>SimulationConstants.m</code>	Sample problem data, <b>not</b> known to the estimator.
<code>CalculateInputs.m</code>	Matlab function used to calculate the input wheel speeds.
<code>Uniform.m</code> ,	Uniform random number generators.
<code>UniformMinMax.m</code>	

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<sup>1</sup>In `script.m` the following one-step method for numerical integration is implemented: Assuming  $v_t(t)$  and  $v_r(t)$  are constant over the sampling interval  $[t_k, t_{k+1}]$  (which they are since piecewise constant inputs are considered in the simulation), we have from (3) for  $t \in [t_k, t_{k+1}]$

$$\dot{r}(t) = v_r \quad \Rightarrow \quad r(t) = r(t_k) + (t - t_k)v_r \quad \Rightarrow \quad r(t_{k+1}) = r(t_k) + (t_{k+1} - t_k)v_r,$$

where  $v_r = v_r(t)$  for all  $t \in [t_k, t_{k+1}]$ . With this and (1), we write for  $x(t)$ ,  $t \in [t_k, t_{k+1}]$

$$\begin{aligned} \dot{x}(t) &= v_t \cos(r(t_k)) + (t - t_k)v_r \\ &= v_t (\cos(r(t_k)) \cos((t - t_k)v_r) - \sin(r(t_k)) \sin((t - t_k)v_r)) \\ &\approx v_t \cos(r(t_k)) - v_t v_r \sin(r(t_k))(t - t_k), \quad \text{for small } (t - t_k) \\ \Rightarrow x(t_{k+1}) &= x(t_k) + v_t \cos(r(t_k))(t_{k+1} - t_k) - \frac{1}{2} v_t v_r \sin(r(t_k))(t_{k+1} - t_k)^2, \end{aligned}$$

where  $v_t = v_t(t)$  for all  $t \in [t_k, t_{k+1}]$ , and similarly for  $y(t)$ ,  $t \in [t_k, t_{k+1}]$

$$y(t_{k+1}) = y(t_k) + v_t \sin(r(t_k))(t_{k+1} - t_k) - \frac{1}{2} v_t v_r \cos(r(t_k))(t_{k+1} - t_k)^2.$$

Clearly, one may also employ other integration schemes such as forward Euler (less accurate) or higher-order methods (e.g. in the Matlab ODE suite; usually slower).

## Task

Implement your solution for the Extended Kalman Filter in the file `Estimator.m`. Your code has to run with the Matlab script `script.m` and problem data as for example given in `PhysicalConstants.m` and `SimulationConstants.m`. For your estimator, use the function definition as given in the template `Estimator.m`.

For evaluating your solution, we will test it on the given problem data. Moreover, we will do suitable modifications of the parameters in `PhysicalConstants.m` and `SimulationConstants.m` and also test your estimator on those.

For judging your own solution, a typical performance of an Extended Kalman Filter implementation for the given problem is shown in Fig. 2 and 3.

## Deliverables

Please hand in by e-mail your implementation of the Extended Kalman Filter in `Estimator.m`. Include the file into a zip-file, which you name `RFE10Ex1_Names.zip`, where *Names* is a list of the **pre- and surnames** of all students<sup>2</sup> who have worked on the solution (for example `RFE10Ex1_AngelaSchoellig_SebastianTrimpe.zip`).

Send your file to Sebastian ([strimpe@ethz.ch](mailto:strimpe@ethz.ch)) until the due date indicated above. We will send a confirmation e-mail upon receiving your solution. You are ultimately responsible that we receive your solution in time.

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<sup>2</sup>Up to three students are allowed to work together on the programming exercise. They will all receive the same grade.

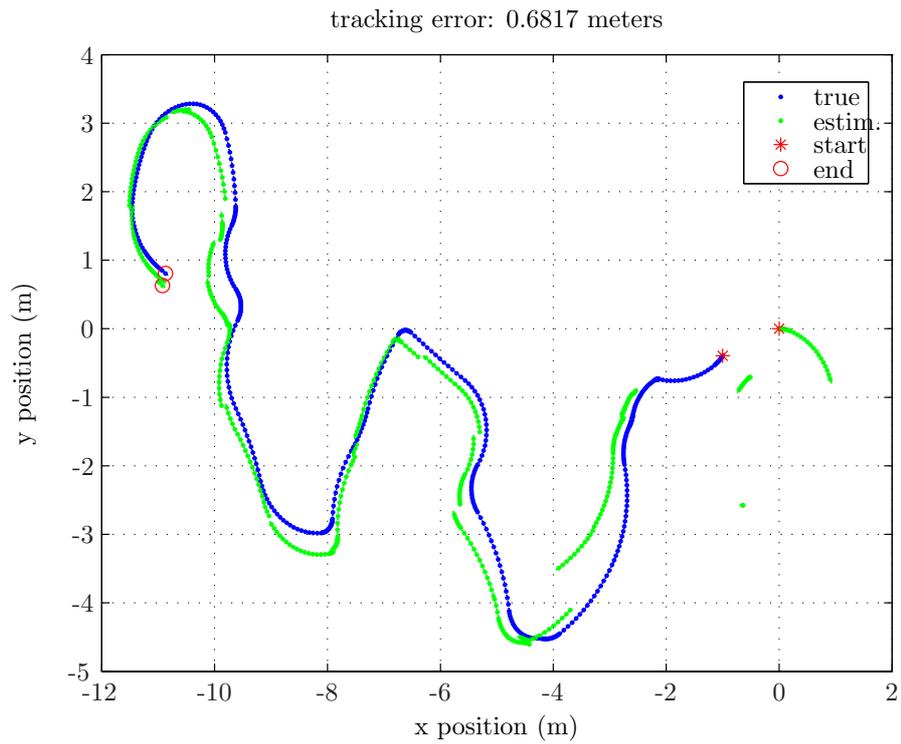


Figure 2: Typical tracking performance of an estimator for the given problem data.

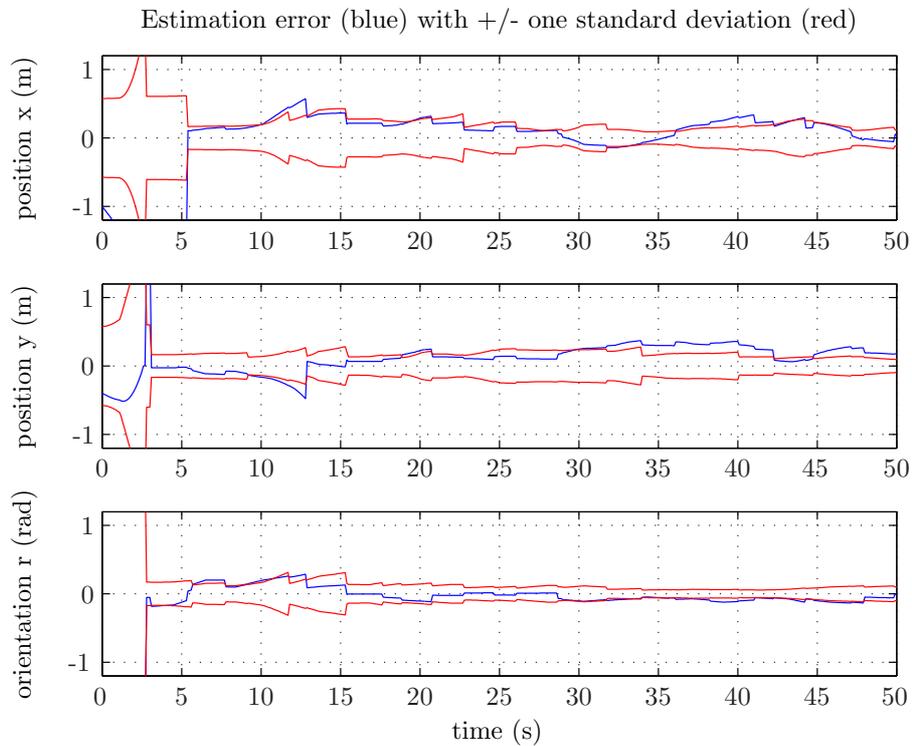


Figure 3: Typical estimation errors with  $\pm$  one standard deviation for the given problem data.