## **HOMEWORK 3**

## 1. About the Kalman filter

Estimating the position of a car. Figure 1 shows the estimation problem using a nonlinear measurement. The process model has been linearized and discretized as,

$$\mathbf{x}_{k} = f(\mathbf{x}_{k-1}, \mathbf{u}_{k-1}, \mathbf{w}_{k-1}) \\ = \begin{bmatrix} 1 & \Delta t \\ 0 & 1 \end{bmatrix} \mathbf{x}_{k-1} + \begin{bmatrix} 0 \\ \Delta t \end{bmatrix} \mathbf{u}_{k-1} + \mathbf{w}_{k-1}$$

The nonlinear measurement equation is,

$$y_k = \phi_k = h(p_k, v_k) = \arctan\left(\frac{S}{D - p_k}\right) + v_k$$

The process noise and measurement noise are assumed to be white noise, i.e.,

$$v_k \sim \mathcal{N}(0, 0.05), \mathbf{w}_k \sim \mathcal{N}(\mathbf{0}, (0.1) \cdot \mathbf{1}_{2 \times 2})$$

The initial state is

$$\mathbf{x}_0 \sim \mathcal{N}\left( \begin{bmatrix} 0\\5 \end{bmatrix}, \begin{bmatrix} 0.01 & 0\\0 & 1 \end{bmatrix} \right)$$

The sample instant is  $\Delta t = 0.5$ s, the initial input is  $u_0 = -2$ m/s<sup>2</sup>, the measurements available are

$$y_1 = 20 \deg, S = 20 m, D = 60 m$$

(1) Try to derive the Extended Kalman filter and Unscented Kalman filter estimate for  $\hat{x}_1$ ,  $\hat{P}_1$ , compare the 2 results.



FIGURE 1. Estimate the position of a car. 1

(2) If the position observation is given as

$$y_k = [1 \ 0] \boldsymbol{x}_k + v_k$$

discuss the existence of a steady-state discrete-time Kalman filter.

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