

Unscented Kalman filter (UKF)

CTRV model

State Vector (x)

$$x = (x, y, v, \psi, \dot{\psi})$$

position (P_x, P_y); Velocity (v); yaw angle (ψ),
yaw rate or turn rate ($\dot{\psi}$)

Prediction : $\begin{cases} \text{Generate sigma points} \\ \text{predict sigma points} \\ \text{predict mean \& covariance} \end{cases}$

update : $\begin{cases} \text{Predict measurement} \\ \text{update state} \end{cases}$

① Generate sigma points

$$X_{k|k} = X_{k|k} : X_{k|k} + \sqrt{(\lambda + n_x) \cdot P_{k|k}}$$

(5x11)

$$X_{k|k} - \sqrt{(\lambda + n_x) P_{k|k}}$$

$$X_{k|k} = \begin{bmatrix} P_x \\ P_y \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

(Object state mean)

$$\lambda = 3 - n_x$$

$$n_x = 5$$

for Initial state

$$\text{no. of sigma points} = 2(n_x) + 1 = 11$$

$P_{K|K}$
 (5x5)
~~predicted~~
 Object state
 covariance
 (Initial state)

$$= \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Augment state

$x_{a,k}$
 (7x1)

$$= \begin{bmatrix} p_x \\ p_y \\ v \\ \varphi \\ \dot{\varphi} \\ \ddot{v}_a \\ \ddot{\varphi} \end{bmatrix}$$

~~process~~
 noise
 vector

$\ddot{v}_a \rightarrow$ longitudinal acceleration noise

$\ddot{\varphi} \rightarrow$ yaw acceleration noise

$n_a = 7$
 No. of sigmas = $2(n_a) + 1$
 points) = 15

Augment
 covariance
 matrix

$P_{a,k|k}$
 (7x7)

$$= \begin{bmatrix} P_{k|k} & 0 \\ 0 & Q \end{bmatrix}$$

(process noise
 covariance
 matrix)

$$Q = \begin{bmatrix} \sigma_a^2 & 0 \\ 0 & \sigma_{\varphi}^2 \end{bmatrix}$$

(2x2)

Generate Augment Sigma points

$$X_{a,k|k} = X_{a,k|k} \cdot \sqrt{(\lambda + n_a) P_{a,k|k}} \quad (7 \times 5)$$

$$X_{a,k|k} = \sqrt{(\lambda + n_a) P_{a,k|k}}$$



$$\lambda = 3 - n_a$$

$$= \text{error}$$

matrix of sigma points
in state space.

② Predict sigma points

$$X_{k+1|k} = f(X_k, V_k)$$

(process model)

$$X_{k+1|k} = X_{a,k|k} + \int_{-t_k}^{t_{k+1}} g(x) dt + V_k$$

$$\tilde{x} = g(x) = \begin{bmatrix} \dot{p}_x \\ \dot{p}_y \\ \dot{v} \\ \dot{\psi} \\ \ddot{\psi} \end{bmatrix} = \begin{bmatrix} \cos \psi \cdot v \\ \sin \psi \cdot v \\ 0 \\ \dot{\psi} \\ 0 \end{bmatrix}$$

$$V_k = \begin{bmatrix} \frac{1}{2} \Delta t^2 \cdot \cos \psi_k \cdot V_a \\ \frac{1}{2} \Delta t^2 \cdot \sin \psi_k \cdot V_a \\ \Delta t \cdot V_a \\ \Delta t^2 \cdot \ddot{\psi} \\ \Delta t \cdot \ddot{\psi} \end{bmatrix}$$

process
noise
Vector

③ Predict Mean & Covariance

a) Predicted state mean.

$$\hat{x}_{k+1|k} = \sum_{i=0}^{2 \cdot n_a} w_i \cdot x_{k+1|k,i}$$

$$w_i = \frac{\lambda}{\lambda + n_a} = \frac{1}{2(\lambda + n_a)} \quad i=0 \dots n_a$$

weights

b) Predicted state Covariance

$$P_{k+1|k} = \sum_{i=0}^{2 \cdot n_a} w_i (x_{k+1|k,i} - \hat{x}_{k+1|k}) \cdot (x_{k+1|k,i} - \hat{x}_{k+1|k})^T$$

④ Predict Measurement

a) $z_{k+1|k} = h(x_{k+1})$
 (3x1) (measurement model)

for LIDAR

$$h(x_{k+1}) = \begin{bmatrix} p_x \\ p_y \end{bmatrix}$$

(2x1)

for RADAR

$$h(x_{k+1}) = \begin{bmatrix} \sqrt{p_x^2 + p_y^2} \\ \arctan(p_y/p_x) \\ \frac{p_x v_x + p_y v_y}{\sqrt{p_x^2 + p_y^2}} \end{bmatrix}$$

(3x1)

$$n_o = 2(n_a) + 1 = 15$$

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if we pass earlier generated sigma points through measurement model to get matrix of sigma points in measurement space

b) Predict Measurement mean

$$\hat{z}_{k+1|k} = \sum_{i=0}^{2-n_o} w_i^o \cdot z_{k+1|k,i}$$

b) Predict Measurement Covariance

$$P_{k+1|k} = \sum_{i=0}^{2-n_o} w_i \left(z_{k+1|k,i} - \hat{z}_{k+1|k} \right) \cdot \left(z_{k+1|k,i} - \hat{z}_{k+1|k} \right)^T + R$$

$$R_{\text{LIDAR}} = \begin{bmatrix} \sigma_R^2 & 0 \\ 0 & \sigma_R^2 \end{bmatrix}$$

$$R_{\text{RADAR}} = \begin{bmatrix} \sigma_r^2 & 0 & 0 \\ 0 & \sigma_\psi^2 & 0 \\ 0 & 0 & \sigma_\psi^2 \end{bmatrix}$$

5) Update state

$$a) T_{k+1|k} = \sum_{i=0}^{2-n} w_i (X_{k+1|k,i} - X_{k+1|k}) (Z_{k+1|k,i} - Z_{k+1|k})^T$$

cross correlation between signal points in state space & measurement space

$$b) K_{k+1|k} = T_{k+1|k} S_{k+1|k}^{-1}$$

Kalman Gain

c) state update

$$X_{k+1|k+1} = X_{k+1|k} + K_{k+1|k} (Z_{k+1|k} - Z_{k+1|k})$$

(5x1)

$$Z_{RADAR} = \begin{bmatrix} P_x \\ P_y \end{bmatrix}$$

$$Z_{RADAR} = \begin{bmatrix} p \\ \varphi \\ \dot{\varphi} \end{bmatrix}$$

d) Covariance matrix update

$$P_{k+1|k+1} = P_{k+1|k} - K_{k+1|k} S_{k+1|k} K_{k+1|k}^T$$

(5x5)

Process - moment ()
{ (1) Initialization ()
(2) Prediction ()

{

- a) Augment-sigma-point ()
- b) Predict-sigma-point ()
- c) Predict mean and Covariance ()

}

(3) update-LIDAR ()
{

- a) Predict Measurement ()
- b) Update State ()

}

(4) update-Radar ()
{

- a) Predict Measurement ()
- b) Update State ()

}

}

Initialize T^0 ()
P

use Initialize (1) state vector z

(2) state covariance matrix

Yaml change.

(1) Initialization ()

(2) Predict-signapoint ()