

EKF Case Study 1 – Vehicle Sensor Fusion

ECE 495/595 Lecture Slides

Winter 2017

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Summary and Quick Links

These slides contain the following concepts:

- ▶ Unicycle observer state space model (Slide 3)
- ▶ Measurement model (Slide 6)



Unicycle Observer Model

- ▶ The eventual goal of the Kalman filter will be to combine various sensor data and produce a complete state estimate.
- ▷ Specifically, the filter will be designed to use the following information:
 - > GPS receiver (x and y position)
 - \rightarrow Wheel encoders (forward speed v)
 - > Gyroscope (yaw rate $\dot{\psi}$)
- ▶ A compass can be very hard to calibrate properly, so it won't be included.
- ▶ Based on this, define the observer state space model.



Unicycle Observer Model

▶ Discrete state equation:

$$\mathbf{X}_{k+1} = f(\mathbf{X}_k) \Rightarrow \begin{cases} x_{k+1} &= x_k + T_s v_k \cos \psi_k \\ y_{k+1} &= y_k + T_s v_k \sin \psi_k \\ \psi_{k+1} &= \psi_k + T_s \dot{\psi}_k \\ v_{k+1} &= v_k \\ \dot{\psi}_{k+1} &= \dot{\psi}_k \end{cases}$$

▷ Observation equation:

$$\mathbf{Z}_k = h(\mathbf{X}_k) \Rightarrow \begin{cases} \operatorname{GPS}_x &= x_k \\ \operatorname{GPS}_y &= y_k \\ \operatorname{Enc.} &= v_k \\ \operatorname{Gyro} &= \dot{\psi}_k \end{cases}$$



Unicycle Observer Model

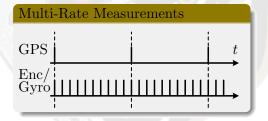
Since the Kalman filter operates on a linear state space model, the state space must be linearized:

$$\mathbf{A}_{k} = \left. \frac{\partial f}{\partial \mathbf{X}} \right|_{k} = \begin{bmatrix} 1 & 0 & -T_{s}v_{k}\sin\psi_{k} & T_{s}\cos\psi_{k} & 0\\ 0 & 1 & T_{s}v_{k}\cos\psi_{k} & T_{s}\sin\psi_{k} & 0\\ 0 & 0 & 1 & 0 & T_{s}\\ 0 & 0 & 0 & 1 & 0\\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{C}_k = \left. \frac{\partial h}{\partial \mathbf{X}} \right|_k = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$



Most GPS receivers update between 1 Hz and 10 Hz, while encoders and inertial sensors update much faster, usually at least 50 Hz:



▶ This means not all measurements are present during every iteration of the Kalman filter.



- ▶ Therefore, the measurement model behind the Kalman filter must be constructed piece-wise.
- ▶ Each measurement will have a separate:
 - > Standard deviation r
 - > Measurement equation $h = f(\mathbf{X}_k)$
 - > Measurement Jacobian row \mathbf{C}_i
- ▶ The complete measurement model is then constructed from all present measurements each time the filter is iterated.
- ▶ The filter sample time should be the same as the fastest sensor.



- ▶ GPS East measurement:
 - > Standard deviation r_{GPS_x}
 - > Measurement equation $h_{\text{GPS}_x} = x_k$
 - > Jacobian row:

$$C_{\text{GPS}_x} = \frac{\partial h_{\text{GPS}_x}}{\partial \mathbf{X}} = \begin{bmatrix} 1 & 0 & 0 & 0 \end{bmatrix}$$

- ▷ GPS North measurement:
 - > Standard deviation r_{GPS_y}
 - > Measurement equation $h_{GPS_u} = y_k$
 - > Jacobian row:

$$C_{\mathrm{GPS}_y} = \frac{\partial h_{\mathrm{GPS}_y}}{\partial \mathbf{X}} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \end{bmatrix}$$



- ▶ Forward speed measurement:
 - > Standard deviation $r_{\rm Enc.}$
 - > Measurement equation $h_{\text{Enc.}} = v_k$
 - > Jacobian row:

$$C_{\text{Enc.}} = \frac{\partial h_{\text{Enc.}}}{\partial \mathbf{X}} = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 \end{bmatrix}$$

- ▷ Gyro measurement:
 - > Standard deviation $r_{\rm Gyro}$
 - > Measurement equation $h_{\rm Gyro} = \dot{\psi}_k$
 - > Jacobian row:

$$C_{\text{Gyro}} = \frac{\partial h_{\text{Gyro}}}{\partial \mathbf{X}} = \begin{bmatrix} 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$



▶ For example, on iterations where only encoder and gyro readings are available, the measurement model will be:

$$\mathbf{R} = \begin{bmatrix} r_{\text{Enc.}}^2 & 0\\ 0 & r_{\text{Gyro}}^2 \end{bmatrix}, \qquad h(\mathbf{X}_k) = \begin{cases} h_{\text{Enc.}} &= v_k\\ h_{\text{Gyro}} &= \dot{\psi}_k \end{cases}$$

$$\mathbf{C}_k = \begin{bmatrix} C_{\text{Enc.}} \\ C_{\text{Gyro}} \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$



▶ On iterations where GPS measurements are also available, the complete measurement model is used:

$$\mathbf{R} = \begin{bmatrix} r_{\text{GPS}_x}^2 & 0 & 0 & 0\\ 0 & r_{\text{GPS}_y}^2 & 0 & 0\\ 0 & 0 & r_{\text{Enc.}}^2 & 0\\ 0 & 0 & 0 & r_{\text{Gyro}}^2 \end{bmatrix}$$

$$h(\mathbf{X}_k) = \begin{cases} h_{\text{GPS}_x} &= x_k \\ h_{\text{GPS}_y} &= y_k \\ h_{\text{Enc.}} &= v_k \\ h_{\text{Gyro}} &= \dot{\psi}_k \end{cases}$$



$$\mathbf{C}_k = \begin{bmatrix} C_{\text{GPS}_x} \\ C_{\text{GPS}_y} \\ C_{\text{Enc.}} \\ C_{\text{Gyro}} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$