

# State space models

Sensor fusion & nonlinear filtering

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# DISCRETE-TIME STATE SPACE MODELS

## Discrete-time state space models

For a **state vector**,  $\mathbf{x}_k$ , and a **measurement vector**,  $\mathbf{y}_k$ , where  $k$  denotes a discrete time index, we have the following models,

*Motion model:*  $\mathbf{x}_k = f_{k-1}(\mathbf{x}_{k-1}, \mathbf{q}_{k-1})$  (1)

*Measurement model:*  $\mathbf{y}_k = h_k(\mathbf{x}_k, \mathbf{r}_k)$  (2)

where  $\mathbf{x}_0 \sim p(\mathbf{x}_0)$ .

- We also assume that both the motion noise,  $\mathbf{q}_{k-1}$ , and the measurement noise,  $\mathbf{r}_k$ , are **independent** of all other noise vectors.

$\leadsto$  this ensures the Markov property (see the next video).

# THE MOTION MODEL

## Motion / process model

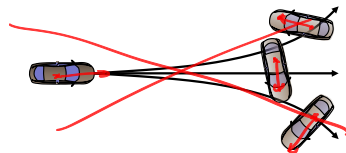
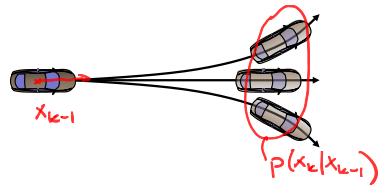
- The system dynamics are described by (1),

$$\mathbf{x}_k = f_{k-1}(\mathbf{x}_{k-1}, \mathbf{q}_{k-1}), \Leftrightarrow p(\mathbf{x}_k | \mathbf{x}_{k-1})$$

which we refer to as the **motion / process model**.

### Note:

- It describes the state evolution,  $p(\mathbf{x}_k | \mathbf{x}_{k-1})$ , i.e., the distribution of  $\mathbf{x}_k$  given  $\mathbf{x}_{k-1}$ .
- The motion model thus connects state over time and helps us to rule out unreasonable trajectories.



# THE MEASUREMENT MODEL

## Measurement model

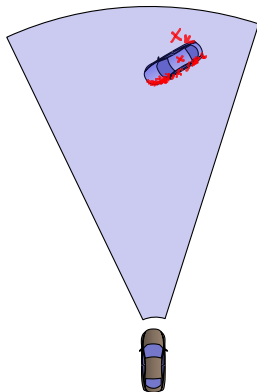
- How the measurements relate to the state vector is described by (2),

$$\mathbf{y}_k = h_k(\mathbf{x}_k, \mathbf{r}_k) \Leftrightarrow p(\mathbf{y}_k | \mathbf{x}_k)$$

and is called the **measurement model** or the **sensor model**.

### Note:

- It describes the distribution of  $\mathbf{y}_k$  given  $\mathbf{x}_k$ ,  $p(\mathbf{y}_k | \mathbf{x}_k)$ , i.e., it defines the **likelihood function**.
- The measurement model relates data to the state vector and helps us to use data to learn about the states.



# MODELS WITH INPUT VARIABLES

## Known input signal

- The system may also have a **known input signal**,  $\mathbf{u}_k$ ,

$$\begin{cases} \mathbf{x}_k = f_{k-1}(\mathbf{x}_{k-1}, \mathbf{u}_k, \mathbf{q}_{k-1}) \\ \mathbf{y}_k = h_k(\mathbf{x}_k, \mathbf{u}_k, \mathbf{r}_k). \end{cases} \iff \begin{cases} p(\mathbf{x}_k | \mathbf{x}_{k-1}; \mathbf{u}_k), \\ p(\mathbf{y}_k | \mathbf{x}_k; \mathbf{u}_k), \end{cases}$$

The time index for  $\mathbf{u}$  in the motion model can also be  $k - 1$ .

- The input signal is often a **control signal** but it may also be an **accurate measurement**.

# SELF-ASSESSMENT

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An important benefit with having both a measurement and a motion model is that past data can provide information about the current state.

- True.
- False.