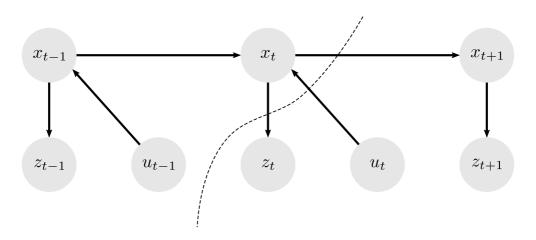
THE KALMAN FILTER



1 DIMENSION

$$x_t = a_t \cdot x_{t-1} + b_t \cdot u_t + \epsilon_{R_t}$$
$$z_t = c_t \cdot x_t + \epsilon_{Q_t}$$

The term ϵ_{R_t} is the system noise:

$$p\left(\epsilon_{R_t}\right) = \mathcal{N}\left(0, \, \sigma_{R_t}^2\right)$$

The term ϵ_{Q_t} is the measurement noise:

$$p\left(\epsilon_{Q_t}\right) = \mathcal{N}\left(0, \sigma_{Q_t}^2\right)$$

1. PREDICTION:

$$p(x_{t} | x_{t-1}, u_{t}) = \mathcal{N}\left(a_{t} \cdot x_{t-1} + b_{t} \cdot u_{t}, \sigma_{R_{t}}^{2}\right)$$

$$bel(x_{t-1}) = p(x_{t-1} | z_{t-1}) = \frac{p(z_{t-1} | x_{t-1}) \cdot p(x_{t-1})}{p(z_{t-1})} = \alpha \cdot p(z_{t-1} | x_{t-1}) \cdot \overline{bel}(x_{t-1}) =$$

$$= \mathcal{N}(\mu_{t-1}, \sigma_{t-1}^{2})$$

$$\overline{bel}(x_{t}) = \int_{x_{t-1} = -\infty}^{+\infty} p(x_{t} | x_{t-1}, u_{t}) \cdot bel(x_{t-1}) \cdot dx_{t-1} =$$

$$= \int_{x_{t-1} = -\infty}^{+\infty} \mathcal{N}\left(a_{t} \cdot x_{t-1} + b_{t} \cdot u_{t}, \sigma_{R_{t}}^{2}\right) \cdot \mathcal{N}\left(\mu_{t-1}, \sigma_{t-1}^{2}\right) \cdot dx_{t-1} = \mathcal{N}\left(\overline{\mu_{t}}, \overline{\sigma_{t}}^{2}\right)$$

$$\overline{\mu_{t}} = a_{t} \cdot \mu_{t-1} + b_{t} \cdot u_{t}$$

$$\overline{\sigma_{t}^{2}} = a_{t}^{2} \cdot \sigma_{t-1}^{2} + \sigma_{R}^{2}.$$

2. CORRECTION:

$$p(z_t | x_t) = \mathcal{N}(c_t \cdot x_t, \sigma_{Q_t}^2)$$

$$\overline{bel}(x_t) = \mathcal{N}(\overline{\mu_t}, \overline{\sigma_t}^2)$$

$$bel(x_t) = p(x_t | z_t) = \alpha \cdot p(z_t | x_t) \cdot \overline{bel}(x_t) =$$

$$= \alpha \cdot \mathcal{N}(c_t \cdot x_t, \sigma_{Q_t}^2) \cdot \mathcal{N}(\overline{\mu_t}, \overline{\sigma_t}^2) = \mathcal{N}(\mu_t, \sigma_t^2)$$

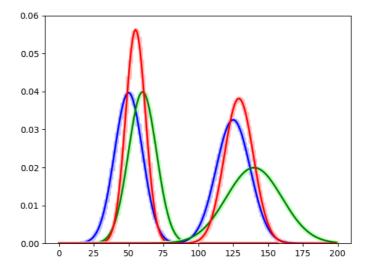
$$K_t = \frac{c_t \cdot \overline{\sigma_t}^2}{c_t^2 \cdot \overline{\sigma_t}^2 + \sigma_{Q_t}^2}$$

$$\mu_t = \overline{\mu_t} + K_t \cdot (z_t - c_t \overline{\mu_t})$$

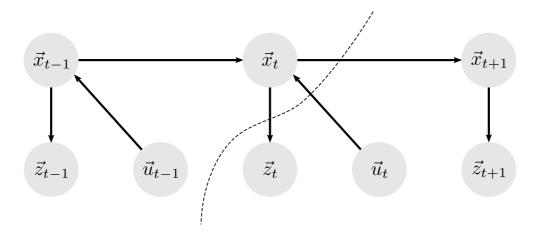
$$\sigma_t^2 = (1 - K_t \cdot c_t) \cdot \overline{\sigma_t}^2$$

$$K_t = 0 \longrightarrow \begin{array}{l} \mu_t = \overline{\mu_t} \\ \sigma_t^2 = \overline{\sigma_t}^2 \end{array}$$

$$K_t > 0 \longrightarrow \sigma_t^2 < \overline{\sigma_t}^2$$



N DIMENSIONS



$$\vec{x}_t = A_t \cdot \vec{x}_{t-1} + B_t \cdot U_t + \epsilon_{R_t}$$
$$\vec{z}_t = C_t \cdot \vec{x}_t + \epsilon_{Q_t}$$

The term ϵ_{R_t} is the system noise:

$$p\left(\epsilon_{R_t}\right) = \mathcal{N}\left(0, R_t\right)$$

The term ϵ_{Q_t} is the measurement noise:

$$p\left(\epsilon_{Q_t}\right) = \mathcal{N}\left(0, Q_t\right)$$

1. PREDICTION:

$$\begin{split} p\left(\vec{x}_{t} \,|\, \vec{x}_{t-1},\, U_{t}\right) &= \mathcal{N}\left(A_{t} \,\cdot\, \vec{x}_{t-1} \,+\, B_{t} \,\cdot\, U_{t},\, R_{t}\right) \\ bel\left(\vec{x}_{t-1}\right) &= p\left(\vec{x}_{t-1} \,|\, \vec{z}_{t-1}\right) = \frac{p\left(\vec{z}_{t-1} \,|\, \vec{x}_{t-1}\right) \,\cdot\, p\left(\vec{x}_{t-1}\right)}{p\left(\vec{z}_{t-1}\right)} \\ &= \alpha \,\cdot\, p\left(\vec{z}_{t-1} \,|\, \vec{x}_{t-1}\right) \,\cdot\, \overline{bel}\left(\vec{x}_{t-1}\right) = \\ &= \mathcal{N}(\vec{\mu}_{t-1},\, \Sigma_{t-1}) \\ \overline{bel}\left(\vec{x}_{t}\right) &= \int_{\vec{x}_{t-1} = -\infty}^{+\infty} p\left(\vec{x}_{t} \,|\, \vec{x}_{t-1},\, U_{t}\right) \,\cdot\, bel\left(\vec{x}_{t-1}\right) \,\cdot\, d\vec{x}_{t-1} = \\ &= \int_{\vec{x}_{t-1} = -\infty}^{+\infty} \mathcal{N}\left(A_{t} \,\cdot\, \vec{x}_{t-1} \,+\, B_{t} \,\cdot\, U_{t},\, R_{t}\right) \,\cdot\, \mathcal{N}\left(\vec{\mu}_{t-1},\, \Sigma_{t-1}\right) \,\cdot\, d\vec{x}_{t-1} = \mathcal{N}\left(\vec{\mu}_{t},\, \overline{\Sigma}_{t}\right) \end{split}$$

$$\vec{\overline{\mu}}_t = A_t \cdot \vec{\mu}_{t-1} + B_t \cdot U_t$$

$$\overline{\Sigma}_t = A_t \cdot \Sigma_{t-1} \cdot A_t^T + R_t$$

2. CORRECTION:

$$\begin{split} p\left(\vec{z}_{t} \mid \vec{x}_{t}\right) &= \mathcal{N}(C_{t} \cdot \vec{x}_{t}, Q_{t}) \\ \overline{bel}\left(\vec{x}_{t}\right) &= \mathcal{N}(\vec{\mu}_{t}, \overline{\Sigma}_{t}) \\ bel\left(\vec{x}_{t}\right) &= p\left(\vec{x}_{t} \mid \vec{z}_{t}\right) = \alpha \cdot p\left(\vec{z}_{t} \mid \vec{x}_{t}\right) \cdot \overline{bel}\left(\vec{x}_{t}\right) = \\ &= \alpha \cdot \mathcal{N}(C_{t} \cdot \vec{x}_{t}, Q_{t}) \cdot \mathcal{N}(\vec{\mu}_{t}, \overline{\Sigma}_{t}) = \mathcal{N}(\vec{\mu}_{t}, \Sigma_{t}) \\ K_{t} &= \overline{\Sigma}_{t} \cdot C_{t}^{T} \cdot \left(C_{t} \cdot \overline{\Sigma}_{t} \cdot C_{t}^{T} + Q_{t}\right)^{-1} \\ \vec{\mu}_{t} &= \vec{\mu}_{t} + K_{t} \cdot \left(\vec{z}_{t} - C_{t} \cdot \overline{\mu}_{t}\right) \\ \Sigma_{t} &= (I - K_{t} \cdot C_{t}) \cdot \overline{\Sigma}_{t} \end{split}$$

$$K_t = 0 \longrightarrow \vec{\mu}_t = \vec{\overline{\mu}}_t$$
$$\Sigma_t = \overline{\Sigma}_t$$

Belonging to \mathbb{R}^{Nx1} : \vec{x}_t , \vec{x}_{t-1} , ϵ_{R_t} , $\vec{\mu}_t$, $\vec{\mu}_{t-1}$, $\vec{\overline{\mu}}_t$.

Belonging to \mathbb{R}^{NxN} : A_t , R_t , $\overline{\Sigma}_t$, Σ_t , Σ_{t-1} .

Belonging to \mathbb{R}^{Mx1} : U_t .

Belonging to \mathbb{R}^{NxM} : B_t .

Belonging to \mathbb{R}^{Lx1} : $\vec{z_t}$, ϵ_{Q_t}

Belonging to \mathbb{R}^{LxN} : C_t .

Belonging to \mathbb{R}^{LxL} : Q_t .

Belonging to \mathbb{R}^{NxL} : K_t