

MCKF Summarize

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The Common KF

Predict:

$$\begin{aligned}\hat{\mathbf{x}}_{k|k-1} &= \mathbf{F}_k \hat{\mathbf{x}}_{k-1|k-1} + \mathbf{B}_k \mathbf{u}_k \\ \mathbf{P}_{k|k-1} &= \mathbf{F}_k \mathbf{P}_{k-1|k-1} \mathbf{F}_k^\top + \mathbf{Q}_k\end{aligned}$$

Update:

$$\begin{aligned}\tilde{\mathbf{y}}_k &= \mathbf{z}_k - \mathbf{H}_k \hat{\mathbf{x}}_{k|k-1} \\ \mathbf{S}_k &= \mathbf{H}_k \mathbf{P}_{k|k-1} \mathbf{H}_k^\top + \mathbf{R}_k \\ \mathbf{K}_k &= \mathbf{P}_{k|k-1} \mathbf{H}_k^\top \mathbf{S}_k^{-1} \\ \hat{\mathbf{x}}_{k|k} &= \hat{\mathbf{x}}_{k|k-1} + \mathbf{K}_k \tilde{\mathbf{y}}_k \\ \mathbf{P}_{k|k} &= (\mathbf{I} - \mathbf{K}_k \mathbf{H}_k) \mathbf{P}_{k|k-1} \\ \tilde{\mathbf{y}}_{k|k} &= \mathbf{z}_k - \mathbf{H}_k \hat{\mathbf{x}}_{k|k}\end{aligned}$$

The MCKF

By fixed point iteration

Predict is exactly same with KF.

Update:

$$\begin{aligned}c &= \hat{\mathbf{x}}(k | k-1) + \tilde{\mathbf{K}}(k)(\mathbf{y}(k) - \mathbf{H}(k)\hat{\mathbf{x}}(k | k-1)) \\ &\text{with} \\ \tilde{\mathbf{K}}(k) &= \tilde{\mathbf{P}}(k | k-1) \mathbf{H}^T(k) \left(\mathbf{H}(k) \tilde{\mathbf{P}}(k | k-1) \mathbf{H}^T(k) + \tilde{\mathbf{R}}(k) \right)^{-1} \\ \tilde{\mathbf{P}}(k | k-1) &= \mathbf{B}_p(k | k-1) \tilde{\mathbf{C}}_x^{-1}(k) \mathbf{B}_p^T(k | k-1) \\ \tilde{\mathbf{R}}(k) &= \mathbf{B}_r(k) \tilde{\mathbf{C}}_y^{-1}(k) \mathbf{B}_r^T(k) \\ \tilde{\mathbf{C}}_x(k) &= \text{diag} (G_\sigma (\tilde{e}_1(k)), \dots, G_\sigma (\tilde{e}_n(k))) \\ \tilde{\mathbf{C}}_y(k) &= \text{diag} (G_\sigma (\tilde{e}_{n+1}(k)), \dots, G_\sigma (\tilde{e}_{n+m}(k))) \\ \tilde{e}_i(k) &= d_i(k) - \mathbf{w}_i(k) \hat{\mathbf{x}}(k | k)_{t-1}\end{aligned}$$

By differential

Predict is exactly same with KF.

Update:

$$\begin{aligned}L_k &= \frac{G_\sigma \left(\|y_k - H \hat{x}_k^- \|_{R_k^{-1}} \right)}{G_\sigma \left(\| \hat{x}_k^- - F \hat{x}_{k-1} \|_{P_{k|k-1}^{-1}} \right)} \\ K_k &= \left(P_{k|k-1}^{-1} + L_k H^T R_k^{-1} H \right)^{-1} L_k H^T R_k^{-1} \\ \hat{x}_k &= \hat{x}_k^- + K_k (y_k - H \hat{x}_k^-) \\ P_{k|k} &= (I - K_k H) P_{k|k-1} (I - K_k H)^T + K_k R_k K_k^T\end{aligned}$$

The UKF

Predict:

$$\begin{aligned}
\mathcal{X}_{t-1} &= (\mu_{t-1} \quad \mu_{t-1} + \gamma\sqrt{\Sigma_{t-1}} \quad \mu_{t-1} - \gamma\sqrt{\Sigma_{t-1}}) \\
\bar{\mathcal{X}}_t^* &= g(u_t, \mathcal{X}_{t-1}) \\
\bar{\mu}_t &= \sum_{i=0}^{2n} w_m^{[i]} \bar{\mathcal{X}}_t^{*[i]} \\
\bar{\Sigma}_t &= \sum_{i=0}^{2n} w_c^{[i]} (\bar{\mathcal{X}}_t^{*[i]} - \bar{\mu}_t) (\bar{\mathcal{X}}_t^{*[i]} - \bar{\mu}_t)^T + Q_t \\
\bar{X}_t &= (\bar{\mu}_t \quad \bar{\mu}_t + \gamma\sqrt{\bar{\Sigma}_t} \quad \bar{\mu}_t - \gamma\sqrt{\bar{\Sigma}_t}) \\
\bar{\mathcal{Z}}_t &= h(\bar{\mathcal{X}}_t) \\
\hat{z}_t &= \sum_{i=0}^{2n} w_m^{[i]} \bar{\mathcal{Z}}_t^{[i]} \\
S_t &= \sum_{i=0}^{2n} w_c^{[i]} (\bar{\mathcal{Z}}_t^{[i]} - \hat{z}_t) (\bar{\mathcal{Z}}_t^{[i]} - \hat{z}_t)^T + R_t
\end{aligned}$$

$\bar{\Sigma}_t$ can also showed as $P^{-1}(k|k-1)$ or $P_x x$, S_t can also showed as P_{zz} .

Update:

$$\begin{aligned}
\bar{\Sigma}_t^{x,z} &= \sum_{i=0}^{2n} w_c^{[i]} (\bar{\mathcal{X}}_t^{[i]} - \bar{\mu}_t) (\bar{\mathcal{Z}}_t^{[i]} - \hat{z}_t)^T \\
K_t &= \bar{\Sigma}_t^{x,z} S_t^{-1} \\
\mu_t &= \bar{\mu}_t + K_t (z_t - \hat{z}_t) \\
\Sigma_t &= \bar{\Sigma}_t - K_t S_t K_t^T
\end{aligned}$$

$\bar{\Sigma}_t^{x,z}$ can also showed as P_{xz}

The MCKF

By fixed point iteration

Predict is exactly same with UKF.

Update:

$$\begin{aligned}
\hat{\mathbf{x}}(k | k)_t &= \hat{\mathbf{x}}(k | k-1) + \tilde{\mathbf{K}}(k)(\mathbf{y}(k) - \hat{\mathbf{y}}(k)) \\
\tilde{\mathbf{K}}(k) &= \tilde{\mathbf{P}}(k | k-1) \mathbf{H}^T(k) \\
&\quad \times \left(\mathbf{H}(k) \tilde{\mathbf{P}}(k | k-1) \mathbf{H}^T(k) + \tilde{\mathbf{R}}(k) \right)^{-1} \\
\tilde{\mathbf{P}}(k | k-1) &= \mathbf{S}_p(k | k-1) \tilde{\mathbf{C}}_x^{-1}(k) \mathbf{S}_p^T(k | k-1) \\
\tilde{\mathbf{R}}(k) &= \mathbf{S}_r(k) \tilde{\mathbf{C}}_y^{-1}(k) \mathbf{S}_r^T(k) \\
\tilde{\mathbf{C}}_x(k) &= \text{diag}(\mathbf{G}_\sigma(\tilde{e}_1(k)), \dots, \mathbf{G}_\sigma(\tilde{e}_n(k))) \\
\tilde{\mathbf{C}}_y(k) &= \text{diag}(\mathbf{G}_\sigma(\tilde{e}_{n+1}(k)), \dots, \mathbf{G}_\sigma(\tilde{e}_{n+m}(k))) \\
\tilde{e}_i(k) &= d_i(k) - \mathbf{w}_i(k) \hat{\mathbf{x}}(k | k)_{t-1}
\end{aligned}$$

where:

$$\begin{aligned}
\mathbf{D}(k) &= \mathbf{S}^{-1}(k) \begin{bmatrix} \hat{\mathbf{x}}(k | k-1) \\ \mathbf{y}(k) - \hat{\mathbf{y}}(k) + \mathbf{H}(k) \hat{\mathbf{x}}(k | k-1) \end{bmatrix} \\
\mathbf{W}(k) &= \mathbf{S}^{-1}(k) \begin{bmatrix} \mathbf{I} \\ \mathbf{H}(k) \end{bmatrix} \\
\mathbf{e}(k) &= \mathbf{S}^{-1}(k) \xi(k) \\
\mathbf{H}(k) &= (\mathbf{P}^{-1}(k | k-1) \mathbf{P}_{\mathbf{xy}}(k))^T
\end{aligned}$$

By differential

Predict is exactly same with UKF.

Update:

$$\begin{aligned}
\mathbf{K}(k) &= (P_{xx}(k) + (P_{zz}(k) - R(k))L^T)^{-1} \tilde{H} R^{-1} \\
x_{k|k} &= \bar{x} + K(k)(y - \bar{y}) \\
P_{k|k} &= (I - K\tilde{H})P_{xx}(I - K\tilde{H})^T + KRK^T
\end{aligned} \tag{1}$$

with:

$$\begin{aligned}
L &= \frac{G(\|y - \tilde{H}\bar{x}\|_{R_k^{-1}})}{G(\|\bar{x} - f(x_{k-1})\|_{P_{xx}^{-1}})} \\
\tilde{H} &= P_{xx}^{-1}P_{xz}
\end{aligned} \tag{2}$$