

Službeni podsjetnik za završni ispit (Teorija estimacije)

Matrični diferencijalni račun

$$\begin{aligned}\frac{\partial(x^T y)}{\partial x} &= y^T, & \frac{\partial(x^T y)}{\partial y} &= x^T \\ \frac{\partial(x^T A x)}{\partial x} &= x^T A^T + x^T A \\ \frac{\partial(Ax)}{\partial x} &= A, & \frac{\partial(x^T A)}{\partial x} &= A \\ \frac{\partial \operatorname{Tr}(ABA^T)}{\partial A} &= AB^T + AB, \\ \frac{\partial \operatorname{Tr}(AB)}{\partial A} &= B^T, & \frac{\partial \operatorname{Tr}(A^T B)}{\partial A} &= B, & \frac{\partial \operatorname{Tr}(BA^T)}{\partial A} &= B\end{aligned}$$

Diskretizacija linearnih sustava

$$\begin{aligned}\Phi &= e^{AT} = I + AT + \frac{A^2 T^2}{2!} + \frac{A^3 T^3}{3!} + \dots = \sum_{v=0}^{\infty} A^v \frac{T^v}{v!} \\ \Gamma &= e^{AT} \left(\int_0^T e^{-A\eta} d\eta \right) B = (e^{AT} - I) A^{-1} B \\ \Phi &\approx I + AT, \quad T \ll \\ \Gamma &\approx BT, \quad T \ll\end{aligned}$$

Nerekurzivni LS estimator

$$\begin{aligned}\hat{x} &= (H^T H)^{-1} H^T y \\ \hat{x} &= (H^T R^{-1} H)^{-1} H^T R^{-1} y\end{aligned}$$

Rekurzivni LS estimator

$$\begin{aligned}K_k &= P_{k-1} H_k^T (R_k + H_k P_{k-1} H_k^T)^{-1} \\ K_k &= P_k H_k^T R_k^{-1} \\ \hat{x}_k &= \hat{x}_{k-1} + K_k (y_k - H_k \hat{x}_{k-1}) \\ P_k &= (I - K_k H_k) P_{k-1} (I - K_k H_k)^T + K_k R_k K_k^T \\ P_k &= (P_{k-1}^{-1} + H_k^T R_k^{-1} H_k)^{-1} \\ P_k &= (I - K_k H_k) P_{k-1}\end{aligned}$$

Diskretni Kalmanov filter

$$\begin{aligned}
x_k &= \Phi_{k-1}x_{k-1} + \Gamma_{k-1}u_{k-1} + w_{k-1} \\
y_k &= H_kx_k + v_k \\
w &\sim (0, Q_k) \\
v &\sim (0, R_k) \\
P_k^- &= \Phi_{k-1}P_{k-1}^+\Phi_{k-1}^T + Q_{k-1} \\
K_k &= P_k^-H_k^T(H_kP_k^-H_k^T + R_k)^{-1} \\
K_k &= P_k^+H_k^TR_k^{-1} \\
\hat{x}_k^- &= \Phi_{k-1}\hat{x}_{k-1}^+ + \Gamma_{k-1}u_{k-1} \\
\hat{x}_k^+ &= \hat{x}_k^- + K_k(y_k - H_k\hat{x}_k^-) \\
P_k^+ &= (I - K_kH_k)P_k^-(I - K_kH_k)^T + K_kR_kK_k^T \\
P_k^+ &= [(P_k^-)^{-1} + H_k^TR_k^{-1}H_k]^{-1} \\
P_k^+ &= (I - K_kH_k)P_k^-
\end{aligned}$$

Kontinuirani Kalmanov filter

$$\begin{aligned}
\dot{x} &= Ax + Bu + w \\
y &= Cx + v \\
w &\sim (0, Q_c) \\
v &\sim (0, R_c) \\
K &= PC^TR_c^{-1} \\
\dot{\hat{x}} &= A\hat{x} + Bu + K(y - C\hat{x}) \\
\dot{P} &= -PC^TR_c^{-1}CP + AP + PA^T + Q_c
\end{aligned}$$

Kontinuirani prošireni Kalmanov filter

$$\begin{aligned}
\dot{x} &= f(x, u, w, t) \\
y &= h(x, v, t) \\
w &\sim (0, Q_c) \\
v &\sim (0, R_c) \\
A &= \left. \frac{\partial f}{\partial x} \right|_{\hat{x}}; \quad L = \left. \frac{\partial f}{\partial w} \right|_{\hat{x}}; \quad C = \left. \frac{\partial h}{\partial x} \right|_{\hat{x}}; \quad M = \left. \frac{\partial h}{\partial v} \right|_{\hat{x}} \\
\tilde{Q}_c &= LQ_cL^T, \quad \tilde{R}_c = MR_cM^T \\
\dot{\hat{x}} &= f(\hat{x}, u, w_0, t) + K[y - h(\hat{x}, v_0, t)] \\
K &= PC^T\tilde{R}_c^{-1} \\
\dot{P} &= -PC^T\tilde{R}_c^{-1}CP + AP + PA^T + \tilde{Q}_c
\end{aligned}$$

Diskretni prošireni Kalmanov filter

$$\begin{aligned}
x_k &= f_{k-1}(x_{k-1}, u_{k-1}, w_{k-1}) \\
y_k &= h_k(x_k, v_k) \\
w &\sim (0, Q_k) \\
v &\sim (0, R_k) \\
\Phi_{k-1} &= \left. \frac{\partial f_{k-1}}{\partial x} \right|_{\hat{x}_{k-1}^+}; \quad L_{k-1} = \left. \frac{\partial f_{k-1}}{\partial w} \right|_{\hat{x}_{k-1}^+}; \quad H_k = \left. \frac{\partial h_k}{\partial x} \right|_{\hat{x}_k^-}; \quad M_k = \left. \frac{\partial h_k}{\partial v} \right|_{\hat{x}_k^-}; \\
\hat{x}_k^- &= f_{k-1}(\hat{x}_{k-1}^+, u_{k-1}, 0) \\
P_k^- &= \Phi_{k-1} P_{k-1}^+ \Phi_{k-1}^T + L_{k-1} Q_{k-1} L_{k-1}^T \\
K_k &= P_k^- H_k^T (H_k P_k^- H_k^T + M_k R_k M_k^T)^{-1} \\
\hat{x}_k^+ &= \hat{x}_k^- + K_k (y_k - h_k(\hat{x}_k^-, 0)) \\
P_k^+ &= (I - K_k H_k) P_k^-
\end{aligned}$$

Iterativni prošireni Kalmanov filter

$$\begin{aligned}
x_k &= f_{k-1}(x_{k-1}, u_{k-1}, w_{k-1}) \\
y_k &= h_k(x_k, v_k) \\
w &\sim (0, Q_k) \\
v &\sim (0, R_k) \\
\Phi_{k-1} &= \left. \frac{\partial f_{k-1}}{\partial x} \right|_{\hat{x}_{k-1}^+}; \quad L_{k-1} = \left. \frac{\partial f_{k-1}}{\partial w} \right|_{\hat{x}_{k-1}^+} \\
\hat{x}_k^- &= f_{k-1}(\hat{x}_{k-1}^+, u_{k-1}, 0) \\
P_k^- &= \Phi_{k-1} P_{k-1}^+ \Phi_{k-1}^T + L_{k-1} Q_{k-1} L_{k-1}^T \\
H_{k,i} &= \left. \frac{\partial h_k}{\partial x} \right|_{\hat{x}_{k,i}^+}; \quad M_{k,i} = \left. \frac{\partial h_k}{\partial v} \right|_{\hat{x}_{k,i}^+}; \\
K_{k,i} &= P_k^- H_{k,i}^T (H_{k,i} P_k^- H_{k,i}^T + M_{k,i} R_k M_{k,i}^T)^{-1} \\
\hat{x}_{k,i+1}^+ &= \hat{x}_k^- + K_{k,i} (y_k - h_k(\hat{x}_{k,i}^+, 0) - H_{k,i} (\hat{x}_k^- - \hat{x}_{k,i}^+)) \\
P_{k,i+1}^+ &= (I - K_{k,i} H_{k,i}) P_k^-
\end{aligned}$$

Informacijski filter

$$\begin{aligned}
x_k &= \Phi_{k-1} x_{k-1} + \Gamma_{k-1} u_{k-1} + w_{k-1} \\
y_k &= H_k x_k + v_k \\
w &\sim (0, Q_k) \\
v &\sim (0, R_k) \\
\mathcal{I}_k^- &= Q_{k-1}^{-1} - Q_{k-1}^{-1} \Phi_{k-1} (\mathcal{I}_{k-1}^+ + \Phi_{k-1}^T Q_{k-1}^{-1} \Phi_{k-1})^{-1} \Phi_{k-1}^T Q_{k-1}^{-1} \\
\mathcal{I}_k^+ &= \mathcal{I}_k^- + H_k^T R_k^{-1} H_k \\
K_k &= (\mathcal{I}_k^+)^{-1} H_k^T R_k^{-1} \\
\hat{x}_k^- &= \Phi_{k-1} \hat{x}_{k-1}^+ + \Gamma_{k-1} u_{k-1} \\
\hat{x}_k^+ &= \hat{x}_k^- + K_k (y_k - H_k \hat{x}_k^-)
\end{aligned}$$

Ustaljeni Kalmanov filter

$$\begin{aligned}
x_k &= \Phi x_{k-1} + \Gamma u_{k-1} + w_{k-1} \\
y_k &= H x_k + v_k \\
w &\sim (0, Q_k) \\
v &\sim (0, R_k) \\
\hat{x}_k^+ &= (I - K_\infty H) \Phi \hat{x}_{k-1}^+ + K_\infty y_k \\
P_\infty &= \Phi P_\infty \Phi^T - \Phi P_\infty H^T (H P_\infty H^T + R)^{-1} H P_\infty \Phi^T + Q \\
K_\infty &= P_\infty H^T (H P_\infty H^T + R)^{-1}
\end{aligned}$$

Alfa-beta filter

$$\begin{aligned}
x_k &= \begin{bmatrix} 1 & T \\ 0 & 1 \end{bmatrix} x_{k-1} + \begin{bmatrix} T^2/2 \\ T \end{bmatrix} w_{k-1} \\
y_k &= \begin{bmatrix} 1 & 0 \end{bmatrix} x_k + v_k \\
w &\sim (0, \sigma_w^2) \\
v &\sim (0, R) \\
K_1 &= -\frac{1}{8}(\lambda^2 + 8\lambda - (\lambda + 4)\sqrt{\lambda^2 + 8\lambda}) = \alpha \\
K_2 &= \frac{1}{4T}(\lambda^2 + 4\lambda - \lambda\sqrt{\lambda^2 + 8\lambda}) = \frac{\beta}{T} \\
\lambda &= \frac{\sigma_w^2 T^2}{R} \\
P_{11}^+ &= K_1 R, \quad P_{12}^+ = K_2 R, \quad P_{22}^+ = \left(\frac{K_1}{T} - \frac{K_2}{2} \right) \frac{K_2 \sigma_w^2}{1 - K_1}
\end{aligned}$$

Diskretni Kalmanov filter za estimaciju sustava s međukoreliranim procesnim i mjernim šumom

$$\begin{aligned}
x_k &= \Phi_{k-1} x_{k-1} + \Gamma_{k-1} u_{k-1} + w_{k-1} \\
y_k &= H_k x_k + v_k \\
w &\sim (0, Q_k) \\
v &\sim (0, R_k) \\
E[w_{k-1} v_j^T] &= M_k \delta_{k-j} \\
\hat{x}_k^- &= \Phi_{k-1} \hat{x}_{k-1}^+ + \Gamma_{k-1} u_{k-1} \\
P_k^- &= \Phi_{k-1} P_{k-1}^+ \Phi_{k-1}^T + Q_{k-1} \\
K_k &= (P_k^- H_k^T + M_k)(H_k P_k^- H_k^T + H_k M_k + M_k^T H_k^T + R_k)^{-1} \\
\hat{x}_k^+ &= \hat{x}_k^- + K_k (y_k - H_k \hat{x}_k^-) \\
P_k^+ &= P_k^- - K_k (H_k P_k^- + M_k^T)
\end{aligned}$$

Diskretni Kalmanov filtar s obojenim mjernim šumom

$$x_k = \Phi_{k-1}x_{k-1} + w_{k-1}$$

$$y_k = H_k x_k + v_k$$

$$v_k = \Psi_{k-1}v_{k-1} + \zeta_{k-1}$$

$$w \sim (0, Q_k)$$

$$\zeta \sim (0, Q_{\zeta,k})$$

$$y'_{k-1} = y_k - \Psi_{k-1}y_{k-1}$$

$$H'_{k-1} = H_k \Phi_{k-1} - \Psi_{k-1} H_{k-1}$$

$$v'_{k-1} = H_k w_{k-1} + \zeta_{k-1}$$

$$y'_k = H'_k x_k + v'_k$$

$$\hat{x}_k^+ = \hat{x}_k^- + K_k(y_k - H_k \hat{x}_k^-)$$

$$\hat{x}_{k+1}^- = \Phi_k \hat{x}_k^+ + C_k(y'_k - H'_k \hat{x}_k^-)$$

$$K_k = P_k^- H_k^{\Gamma} (H'_k P_k^- H_k^{\Gamma} + R_k)^{-1}$$

$$M_k = Q_k H_{k+1}^{\Gamma}$$

$$C_k = M_k (H'_k P_k^- H_k^{\Gamma} + R_k)^{-1}$$

$$P_k^+ = (I - K_k H'_k) P_k^- (I - K_k H'_k)^{\Gamma} + K_k R_k K_k^{\Gamma}$$

$$P_{k+1}^- = \Phi_k P_k^+ \Phi_k + Q_k - C_k M_k^{\Gamma} - \Phi_k K_k M_k^{\Gamma} - M_k K_k^{\Gamma} \Phi_k^{\Gamma}$$