Proof of the Kalman gain matrix: Apriori-Aposteriori form of the discrete Kalman filter

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1 Kalman filter gain matrix

The Kalman filter on apriori-aposteriori form is given by

$$\bar{y}_k = D\bar{x}_k \tag{1}$$

$$\hat{x}_k = \bar{x}_k + K(y_k - \bar{y}_k), \qquad (2)$$

$$\bar{x}_{k+1} = A\hat{x}_k + R_{12}\Delta^{-1}(y_k - \bar{y}),$$
(3)

An alternative derivation of the Kalman filter gain matrix in Eq. (2) is as follows

$$E(\Delta \hat{x}_k \varepsilon_k^T) = E((\Delta \bar{x}_k - K \varepsilon_k) \varepsilon_k^T) = 0.$$
(4)

And from Eq. (4) we have

$$E(\Delta \hat{x}_k \varepsilon_k^T) = E((\Delta \bar{x}_k - K \varepsilon_k) \varepsilon_k^T) = E(\Delta \bar{x}_k \underbrace{(\Delta \bar{x}_k^T D^T + w_k^T)}_{= \bar{X} D^T - K E(\varepsilon_k \varepsilon_k^T) = 0,} - K E(\varepsilon_k \varepsilon_k^T) = 0,$$
(5)

since $E(\Delta \bar{x}_k w_k^T) = 0.$

We then get from Eq. (5) that the optimal Kalman filter gain matrix in the filter is given by

$$K = \bar{X}D^{T}(D\bar{X}D^{T} + W)^{-1},$$
(6)

where we have used that

$$\Delta = \mathcal{E}(\varepsilon_k \varepsilon_k^T) = D\bar{X}D^T + W.$$
(7)

From lecture notes Ch. 2.6.3 main_estim_e2.pdf