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STOCHASTIC ESTIMATION

The Steady-State Kalman-Bucy Filter:
Continuous-Time Case

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MOTIVATION

We seek special properties of continuous-time Kalman-Bucy filters where

- state dynamics are linear and time-invariant
- measurement equation is linear and time-invariant
- the noise statistics are stationary

Optimal only in
statistical steady-state
(effects of initial state
gone).

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SUMMARY OF RESULTS

- 1) As $t \rightarrow \infty$, the error covariance matrix $\underline{\Sigma}(t) \rightarrow \underline{\Sigma} = \text{constant}$
- 2) The filter gain matrix $\underline{H}(t) \rightarrow \underline{H} = \text{constant}$
- 3) The Kalman-Bucy filter is linear and time-invariant; it can be constructed using purely analog elements (integrators, constant gains, summers).

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PROBLEM FORMULATION

- State Dynamics (LTI)

$$\dot{\underline{x}}(t) = \underline{A}\underline{x}(t) + \underline{B}\underline{u}(t) + \underline{L}\underline{\xi}(t) \quad (1)$$

- Measurement Equation

$$\underline{z}(t) = \underline{C}\underline{x}(t) + \underline{e}(t) \quad (2)$$

- \underline{A} , \underline{B} , \underline{L} , \underline{C} constant known matrices

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INITIAL STATE $\underline{x}(0)$

- $\underline{x}(0)$ is gaussian random vector

$$E \{ \underline{x}(0) \} = \bar{\underline{x}}_0 \quad (3)$$

$$\text{cov} [\underline{x}(0); \underline{x}(0)] = \underline{\Sigma}_0 = \underline{\Sigma}'_0 \geq \underline{0} \quad (4)$$

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PLANT NOISE $\underline{\xi}(t)$

$\underline{\xi}(t)$ is stationary, gaussian, white noise

$$E \{ \underline{\xi}(t) \} = \underline{0} \quad (5)$$

$$\text{cov} [\underline{\xi}(t); \underline{\xi}(\tau)] = \underline{\Xi} \delta(t-\tau) \quad (6)$$

$$\underline{\Xi} = \underline{\Xi}' > \underline{0} \quad (\underline{\Xi} = \text{constant matrix}) \quad (7)$$

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MEASUREMENT NOISE $\underline{\theta}(t)$

- $\underline{\theta}(t)$ is stationary, gaussian, white noise

$$E \{ \underline{\theta}(t) \} = \underline{0} \quad (8)$$

$$\text{cov} [\underline{\theta}(t); \underline{\theta}(\tau)] = \underline{\Theta} \delta(t-\tau) \quad (9)$$

$$\underline{\Theta} = \underline{\Theta}' > \underline{0} \quad (\underline{\Theta} = \text{constant matrix}) \quad (10)$$

- $\underline{x}(0), \underline{\xi}(t), \underline{\theta}(\tau)$ mutually independent for all t, τ

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ADDITIONAL ASSUMPTIONS

- Controllability

$$[\underline{A}, \underline{L}] \text{ is controllable pair} \quad (11)$$

- Observability

$$[\underline{A}, \underline{C}] \text{ is observable pair} \quad (12)$$

Recall:

$$\begin{aligned} \dot{\underline{x}}(t) &= \underline{A} \underline{x}(t) + \underline{B} \underline{u}(t) + \underline{L} \underline{\xi}(t) \\ \underline{z}(t) &= \underline{C} \underline{x}(t) + \underline{\theta}(t) \end{aligned}$$

$\underline{\xi}(t)$ must excite all modes

all modes must be observed in $\underline{z}(t)$

$$* \text{rank} [\underline{A} : \underline{A} \underline{L} : \underline{A}^2 \underline{L} : \dots : \underline{A}^{n-1} \underline{L}] = n$$

$$** \text{rank} [\underline{C}' : \underline{A}'^2 \underline{C}' : \dots : \underline{A}'^{n-1} \underline{C}'] = n$$

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TIME VARYING KALMAN-BUCY FILTER

The error covariance matrix $\underline{\Sigma}(t)$ is solution of matrix Riccati differential equation

$$\frac{d}{dt} \underline{\Sigma}(t) = \underline{A} \underline{\Sigma}(t) + \underline{\Sigma}(t) \underline{A}' + \underline{L} \underline{\Xi} \underline{L}' - \underline{\Sigma}(t) \underline{C}' \underline{\Theta}^{-1} \underline{C} \underline{\Sigma}(t) \quad (13)$$

$$\underline{\Sigma}(0) = \text{cov} [\underline{x}(0); \underline{x}(0)] \quad (14)$$

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• The filter gain matrix $\underline{H}(t)$ is

$$\underline{H}(t) = \underline{\Sigma}(t) \underline{C}' \underline{\Theta}^{-1} \quad (15)$$

• The state estimate $\hat{\underline{x}}(t)$ is generated by

$$\frac{d}{dt} \hat{\underline{x}}(t) = \underline{A} \hat{\underline{x}}(t) + \underline{B} u(t) + \underline{H}(t) [\underline{z}(t) - \underline{C} \hat{\underline{x}}(t)] \quad (16)$$

$$\hat{\underline{x}}(0) = E\{\underline{x}(0)\} \quad (17)$$

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ASYMPTOTIC PROPERTIES $t \rightarrow \infty$

The error covariance matrix $\underline{\Sigma}(t)$ approaches a constant limiting value $\underline{\Sigma}$

$$\lim_{t \rightarrow \infty} \underline{\Sigma}(t) = \underline{\Sigma} \quad (18)$$

and, hence,

$$\lim_{t \rightarrow \infty} \underline{H}(t) = \lim_{t \rightarrow \infty} \underline{\Sigma}(t) \underline{C}' \underline{\Theta}^{-1} = \underline{H} \quad (19)$$

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The steady state error covariance matrix $\underline{\Sigma}$ is an unique, positive definite, symmetric solution matrix of the algebraic matrix Riccati equation

$$0 = \underline{A} \underline{\Sigma} + \underline{\Sigma} \underline{A}' + \underline{L} \underline{\Xi} \underline{L}' - \underline{\Sigma} \underline{C}' \underline{\Theta}^{-1} \underline{C} \underline{\Sigma} \quad (20)$$

↑ many solution matrices possible!

In MATLAB the "correct" $\underline{\Sigma}$ is computed directly.

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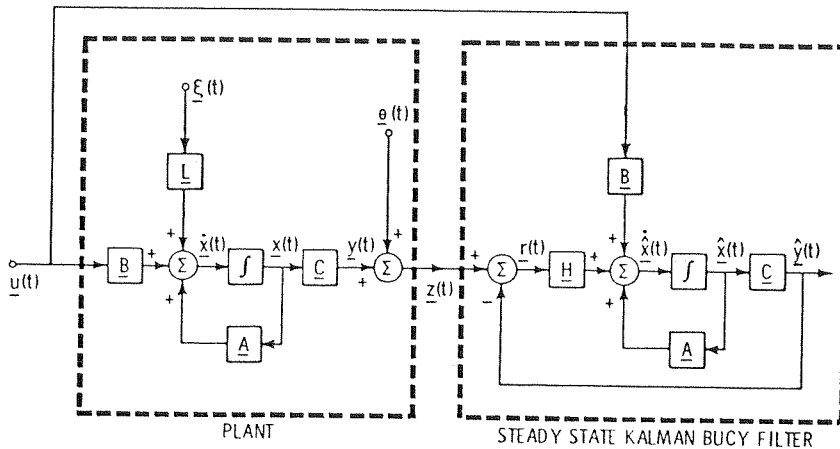
- The steady-state Kalman-Bucy filter generates an estimate $\hat{\underline{x}}(t)$ by

$$\frac{d}{dt} \hat{\underline{x}}(t) = \underline{A} \hat{\underline{x}}(t) + \underline{B} \underline{u}(t) + \underline{H} [z(t) - \underline{C} \hat{\underline{x}}(t)] \quad (21)$$

$$\hat{\underline{x}}(0) = E \{ \underline{x}(0) \} \quad (22)$$

NOTE: The filter (21) can be realized with standard analog hardware

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PROPERTIES

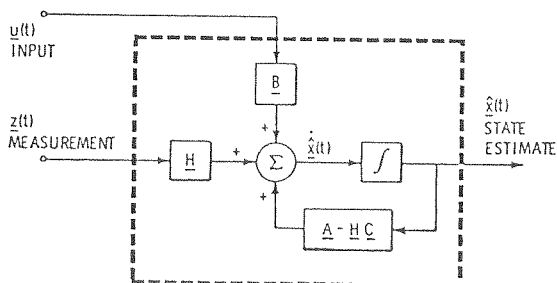
- Alternate form of steady state Kalman-Bucy filter, eq (21)

$$\frac{d}{dt} \hat{\underline{x}}(t) = [\underline{A} - \underline{H}\underline{C}] \hat{\underline{x}}(t) + \underline{B}\underline{u}(t) + \underline{H}\underline{z}(t) \quad (23)$$

- Filter "system" matrix $[\underline{A} - \underline{H}\underline{C}]$ is automatically strictly stable, i.e. all of the eigenvalues of $[\underline{A} - \underline{H}\underline{C}]$ have negative real parts

• Consequence of controllability and observability assumptions.
 • \underline{A} can be unstable!

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RESIDUAL PROPERTY

- As $t \rightarrow \infty$, the residual or innovations process $\underline{r}(t)$

$$\underline{r}(t) \triangleq \underline{z}(t) - \underline{C}\hat{\underline{x}}(t) \quad (24)$$

is gaussian, stationary, zero mean continuous white noise

$$\text{cov} [\underline{r}(t); \underline{r}(\tau)] = \underline{Q} \delta(t - \tau) \quad (25)$$

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SPECIAL CASE: NO PLANT NOISE

- If

$$\underline{\Xi}(t) = \underline{0} \quad (\text{set } \underline{\Xi} = \underline{0} \text{ or } \underline{L} = \underline{0}) \quad (26)$$

then

$$\dot{\underline{\Sigma}}(t) = \underline{A} \underline{\Sigma}(t) + \underline{\Sigma}(t) \underline{A}' - \underline{\Sigma}(t) \underline{C}' \underline{Q}^{-1} \underline{C} \underline{\Sigma}(t) \quad (27)$$

$$\lim_{t \rightarrow \infty} \underline{\Sigma}(t) = \underline{\Sigma} = \underline{0} \quad (28)$$

The only uncertainty is due to initial state

← Time-varying KF is optimal!

(27) $\dot{\underline{\Sigma}}$ equation stable, even if \underline{A} is unstable.

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- This implies that we have perfect estimation

$$\lim \hat{\underline{x}}(t) = \underline{x}(t) \quad (29)$$

- $t \rightarrow \infty$

Note that

$$\lim_{t \rightarrow \infty} \underline{H}(t) = \lim_{t \rightarrow \infty} \underline{\Sigma}(t) \underline{C}' \underline{Q}^{-1} = \underline{H} = \underline{0} \quad (30)$$

i. e. filters runs "open-loop"

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⇒ It makes no sense to have steady state Kalman-Bucy filter if there is no plant noise

- It makes sense to include $\underline{\Xi} > \underline{0}$ (fictitious plant noise) to prevent filter divergence and compensate for plant parameter uncertainties.