

REF. NO. KF # 4

1

STOCHASTIC ESTIMATION

Response of Linear Systems to
White Noise Inputs: Continuous
Time Case

2

PROBLEM FORMULATION

- State Dynamics:

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

deterministic input

continuous-time WN

- Initial State $\underline{x}(t_0)$: random vector
- $\underline{\xi}(t)$: continuous white noise
- $\underline{A}(t)$, $\underline{B}(t)$, $\underline{L}(t)$, $\underline{u}(t)$: deterministic

3

BASIC QUESTIONS

- How does $\underline{x}(t)$ propagate on the average?
→ Determine $E \{ \underline{x}(t) \} \triangleq \underline{\bar{x}}(t)$
- How variable is $\underline{x}(t)$ about its average?
→ Determine $\text{cov} [\underline{x}(t); \underline{x}(t)] \triangleq \underline{\Sigma}(t)$

4

INITIAL UNCERTAINTY

- $\underline{x}(t_0)$ is random vector

$$\underline{x}(t_0) \in R_n \quad (2)$$

$$E \{ \underline{x}(t_0) \} = \bar{\underline{x}}(t_0) \text{ assumed known} \quad (3)$$

$$\text{cov} [\underline{x}(t_0); \underline{x}(t_0)] \quad (4)$$

$$= E \left\{ \left(\underline{x}(t_0) - \bar{\underline{x}}(t_0) \right) \left(\underline{x}(t_0) - \bar{\underline{x}}(t_0) \right)' \right\}$$

$$= \underline{\Sigma}_0 \quad \text{assumed known}$$

$$\underline{\Sigma}_0 = \underline{\Sigma}_0' \geq 0 \quad (5)$$

$\underline{\Sigma}_0$ = initial state uncertainty

5

DRIVING WHITE NOISE

$\underline{\xi}(t)$ white continuous-time noise

$$E \{ \underline{\xi}(t) \} \triangleq \bar{\underline{\xi}}(t) \quad (6)$$

$$\text{cov} [\underline{\xi}(t); \underline{\xi}(\tau)] = \underline{\Xi}(t) \delta(t - \tau) \quad (7)$$

$$\underline{\Xi}(t) = \underline{\Xi}'(t) \geq 0 \quad (8)$$

$\underline{\xi}(t), \underline{x}(t_0)$ independent for all t

$$\rightarrow \text{cov} [\underline{\xi}(t); \underline{x}(t_0)] = 0 \quad (9)$$

Dirac "delta" function



infinite variance
no time-correlation

$\bar{\underline{\xi}}(t), \underline{\Xi}(t)$ assumed known.

6

← called intensity matrix

SUMMARY OF RESULTS

- State $\underline{x}(t) \in R_n$ is a colored Markov process (not white).

- Mean Dynamics

$$E \{ \underline{x}(t) \} \triangleq \bar{\underline{x}}(t) \quad (10)$$

$$\boxed{\begin{aligned} \frac{d}{dt} \bar{\underline{x}}(t) &= \underline{A}(t) \bar{\underline{x}}(t) + \underline{B}(t) \underline{u}(t) \\ &+ \underline{L}(t) \bar{\underline{\xi}}(t) \end{aligned}} \quad (11)$$

$$\bar{\underline{x}}(t_0) = E \{ \underline{x}(t_0) \} \text{ known} \quad (12)$$

7

- $\bar{x}(t)$ is computed either by direct numerical solution of (11) or by the formula

$$\begin{aligned} \bar{x}(t) = & \underline{\Phi}(t, t_0) \bar{x}(t_0) \\ & + \int_{t_0}^t \underline{\Phi}(t, \tau) \underline{B}(\tau) \underline{u}(\tau) d\tau \\ & + \int_{t_0}^t \underline{\Phi}(t, \tau) \underline{L}(\tau) \bar{\xi}(\tau) d\tau \end{aligned} \quad (13)$$

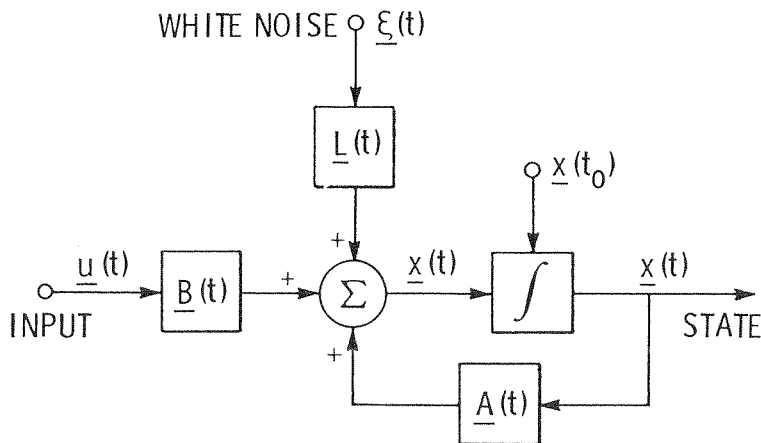
• This is standard convolution for LTV systems

• $\underline{\Phi}(t, \tau)$ is fundamental transition matrix

$$\begin{aligned} \frac{d}{dt} \underline{\Phi}(t, \tau) &= \underline{A}(t) \underline{\Phi}(t, \tau) \\ \underline{\Phi}(\tau, \tau) &= \underline{I} \end{aligned}$$

8

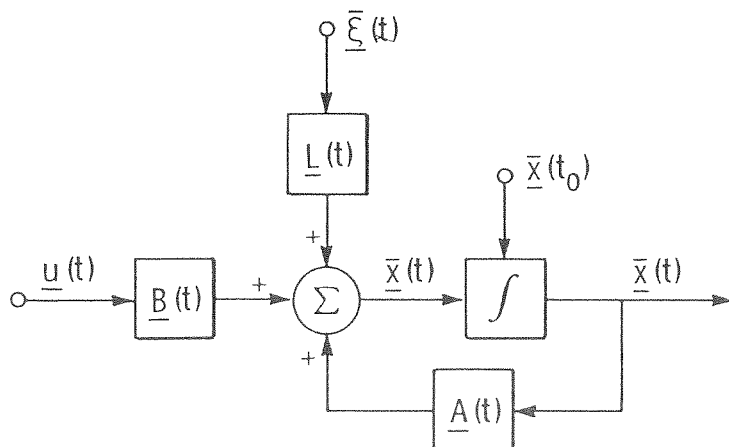
STOCHASTIC DYNAMIC SYSTEM



9

DYNAMIC SYSTEM THAT GENERATES MEAN STATE

Replica of system dynamics



10

• Covariance Dynamics

$$\underline{\Sigma}(t) \triangleq E \left\{ \left(\underline{x}(t) - \bar{\underline{x}}(t) \right) \left(\underline{x}(t) - \bar{\underline{x}}(t) \right)' \right\} \quad (14)$$

$$= \text{cov} \left[\underline{x}(t); \underline{x}(t) \right]$$

$$\underline{\Sigma}(t) = \underline{\Sigma}'(t) \geq \underline{0}$$

$$\boxed{\begin{aligned} \frac{d}{dt} \underline{\Sigma}(t) &= \underline{A}(t) \underline{\Sigma}(t) + \underline{\Sigma}(t) \underline{A}'(t) \\ &\quad + \underline{L}(t) \underline{\Xi}(t) \underline{L}'(t) \end{aligned}}$$

$$\underline{\Sigma}(t_0) = \underline{\Sigma}_0 = \text{cov} \left[\underline{x}(t_0); \underline{x}(t_0) \right]$$

independent of $\underline{B}(t), \underline{u}(t)$!

(15)

(16)

Dynamic evolution of covariance matrix



(17)

LTV matrix differential eq. of Lyapunov type

11

$\underline{\Sigma}(t)$ is determined either from direct numerical solution of Eq (16) or by the formula

$$\begin{aligned} \underline{\Sigma}(t) &= \underline{\Phi}(t, t_0) \underline{\Sigma}(t_0) \underline{\Phi}'(t, t_0) \\ &\quad + \int_{t_0}^t \underline{\Phi}(t, \tau) \underline{L}(\tau) \underline{\Xi}(\tau) \underline{L}'(\tau) \underline{\Phi}'(t, \tau) d\tau \end{aligned} \quad (18)$$

12

• Cross Covariance Dynamics (Correlation Matrix)

$$\begin{aligned} \underline{\Sigma}(t, \tau) &\triangleq E \left\{ \left(\underline{x}(t) - \bar{\underline{x}}(t) \right) \left(\underline{x}(\tau) - \bar{\underline{x}}(\tau) \right)' \right\} \\ &= \text{cov} \left[\underline{x}(t); \underline{x}(\tau) \right]; \quad \tau < t \end{aligned} \quad (19)$$

• Define

$$\underline{\Sigma}(t, t) \triangleq \underline{\Sigma}(t)$$

• Given $\underline{\Sigma}(\tau, \tau)$, $\underline{\Sigma}(t, \tau)$ is computed by

$$\boxed{\underline{\Sigma}(t, \tau) = \underline{\Sigma}(\tau, \tau) \underline{\Phi}'(t, \tau); \quad t \geq \tau} \quad (21)$$

$$\begin{aligned} \frac{d}{dt} \underline{\Phi}(t, \tau) &= \underline{A}(t) \underline{\Phi}(t, \tau) \\ \underline{\Phi}(\tau, \tau) &= \underline{I} \end{aligned} \quad (20)$$

13

DIGRESSION

For proofs we need certain properties of expectation operator

$$\begin{aligned} E \{ \underline{A}(t) \underline{x}(t) + \underline{B}(t) \underline{y}(t) \} \\ = \underline{A}(t) E \{ \underline{x}(t) \} + \underline{B}(t) E \{ \underline{y}(t) \} \end{aligned} \quad (22)$$

14

$$\frac{d}{dt} E \{ \underline{x}(t) \} = E \left\{ \frac{d}{dt} \underline{x}(t) \right\}$$

$$\int_a^b E \{ \underline{x}(t) \} dt = E \left\{ \int_a^b \underline{x}(t) dt \right\}$$

(23) Expectation operator commutes with differentiation and integration operators

(24)

15

DERIVATION OF MEAN DYNAMICS

• State Dynamics

$$\dot{\underline{x}}(t) = \underline{A}(t) \underline{x}(t) + \underline{B}(t) \underline{u}(t) + \underline{L}(t) \underline{\xi}(t) \quad (25)$$

$$E \{ \dot{\underline{x}}(t) \} = \frac{d}{dt} E \{ \underline{x}(t) \} \quad (26)$$

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

• Therefore

$$\boxed{\dot{\bar{\underline{x}}}(t) = \underline{A}(t) \bar{\underline{x}}(t) + \underline{B}(t) \underline{u}(t) + \underline{L}(t) \bar{\underline{\xi}}(t)} \quad (27)$$

16

DEFINITION OF "ERROR" VECTOR $\tilde{\underline{x}}(t)$ (prediction error)

$$\begin{aligned} \tilde{\underline{x}}(t) &\triangleq \underline{x}(t) - \bar{\underline{x}}(t) \\ E \{ \tilde{\underline{x}}(t) \} &= \underline{0} \end{aligned} \quad (28)$$

NOTE: From Eqs (25) and (27)

$$\boxed{\dot{\tilde{\underline{x}}}(t) = \underline{A}(t) \tilde{\underline{x}}(t) + \underline{L}(t) \tilde{\underline{\xi}}(t)} \quad (29)$$

17

$$\tilde{\underline{\xi}}(t) \triangleq \underline{\xi}(t) - \bar{\underline{\xi}}(t) \quad (30)$$

$$E \{ \tilde{\underline{\xi}}(t) \} = \underline{0} \quad (31)$$

$$\begin{aligned} \text{cov} [\tilde{\underline{\xi}}(t); \tilde{\underline{\xi}}(\tau)] &= E \{ \tilde{\underline{\xi}}(t) \tilde{\underline{\xi}}'(\tau) \} \\ &= \text{cov} [\underline{\xi}(t); \underline{\xi}(\tau)] = \underline{\Sigma}(t) \delta(t - \tau) \end{aligned} \quad (32)$$

18

• Hence

$$\begin{aligned} \underline{\Sigma}(t) &= \text{cov} [\underline{x}(t); \underline{x}(t)] \\ &= E \left\{ \left(\underline{x}(t) - \bar{\underline{x}}(t) \right) \left(\underline{x}(t) - \bar{\underline{x}}(t) \right)' \right\} \\ &= E \{ \tilde{\underline{x}}(t) \tilde{\underline{x}}'(t) \} \end{aligned} \quad (33)$$

19

DERIVATION OF COVARIANCE DYNAMICS

$$\begin{aligned} \frac{d}{dt} \underline{\Sigma}(t) &= \frac{d}{dt} E \{ \tilde{\underline{x}}(t) \tilde{\underline{x}}'(t) \} \\ &= E \left\{ \frac{d}{dt} \left(\tilde{\underline{x}}(t) \tilde{\underline{x}}'(t) \right) \right\} \\ &= E \left\{ \dot{\tilde{\underline{x}}}(t) \tilde{\underline{x}}'(t) + \tilde{\underline{x}}(t) \dot{\tilde{\underline{x}}}'(t) \right\} \end{aligned} \quad (34)$$

20

• From Eqs (29) and (34)

$$\begin{aligned} \dot{\underline{\Sigma}}(t) &= E \left\{ \underline{A}(t) \tilde{\underline{x}}(t) \tilde{\underline{x}}'(t) + \underline{L}(t) \tilde{\underline{\xi}}(t) \tilde{\underline{x}}'(t) \right. \\ &\quad \left. + \tilde{\underline{x}}(t) \tilde{\underline{x}}'(t) \underline{A}'(t) + \tilde{\underline{x}}(t) \tilde{\underline{\xi}}'(t) \underline{L}'(t) \right\} \\ &= \underline{A}(t) E \left\{ \underbrace{\tilde{\underline{x}}(t) \tilde{\underline{x}}'(t)}_{\underline{\Sigma}(t)} \right\} + E \left\{ \underbrace{\tilde{\underline{x}}(t) \tilde{\underline{x}}'(t)}_{\underline{\Sigma}(t)} \right\} \underline{A}'(t) \\ &\quad + \underline{L}(t) E \left\{ \tilde{\underline{\xi}}(t) \tilde{\underline{x}}'(t) \right\} + E \left\{ \tilde{\underline{x}}(t) \tilde{\underline{\xi}}'(t) \right\} \underline{L}'(t) \end{aligned} \quad (35)$$

• Need to calculate

$$E \{ \tilde{\underline{\xi}}(t) \tilde{\underline{x}}'(t) \} = \left[E \{ \tilde{\underline{x}}(t) \tilde{\underline{\xi}}'(t) \} \right]' \quad (36)$$

21

$$E \{ \tilde{\underline{x}}(t) \tilde{\underline{x}}'(t) \} \quad (37)$$

$$= E \left\{ \tilde{\underline{x}}(t) \left[\underline{\Phi}(t, t_0) \tilde{\underline{x}}(t_0) + \int_{t_0}^t \underline{\Phi}(t, \tau) \underline{L}(\tau) \tilde{\underline{z}}(\tau) d\tau \right]' \right\}$$

$$= E \left\{ \underbrace{\tilde{\underline{x}}(t) \tilde{\underline{x}}'(t_0)}_0 \right\} \underline{\Phi}'(t, t_0) \quad \leftarrow \text{because initial state is independent of WN}$$

$$+ E \left\{ \tilde{\underline{x}}(t) \int_{t_0}^t \tilde{\underline{z}}(\tau) \underline{L}'(\tau) \underline{\Phi}'(t, \tau) d\tau \right\}$$

22

$$= \int_{t_0}^t E \{ \tilde{\underline{x}}(t) \tilde{\underline{z}}'(\tau) \} \underline{L}'(\tau) \underline{\Phi}'(t, \tau) d\tau$$

$$= \int_{t_0}^t \underline{\Xi}(t) \delta(t-\tau) \underline{L}'(\tau) \underline{\Phi}'(t, \tau) d\tau$$

$$= \frac{1}{2} \underline{\Xi}(t) \underline{L}'(t) \underbrace{\underline{\Phi}'(t, t)}_1 = \frac{1}{2} \underline{\Xi}(t) \underline{L}'(t)$$

Digression

$$\int_a^b f(\tau) \delta(\tau-x) d\tau = f(x) \quad a < x < b$$

$$\int_a^b f(\tau) \delta(\tau-b) d\tau = \frac{1}{2} f(b)$$

23

• From (35) and (37)

$$\dot{\underline{\Sigma}}(t) = \underline{A}(t) \underline{\Sigma}(t) + \underline{\Sigma}(t) \underline{A}'(t) \quad (38)$$

$$+ \frac{1}{2} \underline{L}(t) \underline{\Xi}(t) \underline{L}'(t)$$

$$+ \left[\frac{1}{2} \underline{\Xi}(t) \underline{L}'(t) \right]' \underline{L}'(t)$$

or

$$\boxed{\begin{aligned} \dot{\underline{\Sigma}}(t) &= \underline{A}(t) \underline{\Sigma}(t) + \underline{\Sigma}(t) \underline{A}'(t) \\ &+ \underline{L}(t) \underline{\Xi}(t) \underline{L}'(t) \end{aligned}}$$

LTV Lyapunov matrix DE

(39)

independent of $\underline{B}(t), \underline{u}(t)$

24

STEADY STATE CONSIDERATIONS• Time Invariant State Dynamics

$$\underline{x}(t) = \underline{A} \underline{x}(t) + \underline{B} \underline{u}(t) + \underline{L} \underline{\xi}(t) \quad (40)$$

• Stationary White Noise

$$\text{cov} [\underline{\xi}(t); \underline{\xi}(\tau)] = \underline{\Xi} \delta(t - \tau) \quad (41)$$

- Is there a limiting behavior of $\underline{\Sigma}(t)$?

constant intensity matrix

25

RESULT

- If $\underline{A}$ is strictly stable matrix, i.e.

$$\text{Re } \lambda_i(\underline{A}) < 0; \quad i = 1, 2, \dots, n \quad (42)$$

- Then for all $\underline{\Sigma}(t_0)$

$$\lim_{t \rightarrow \infty} \underline{\Sigma}(t) = \underline{\Sigma} \quad \text{exists} \quad (43)$$

$$\underline{\Sigma} = \underline{\Sigma}' \geq \underline{0} = \text{constant} \quad (44)$$

$$\underline{0} = \underline{A} \underline{\Sigma} + \underline{\Sigma} \underline{A}' + \underline{L} \underline{\Xi} \underline{L}' \quad (45)$$

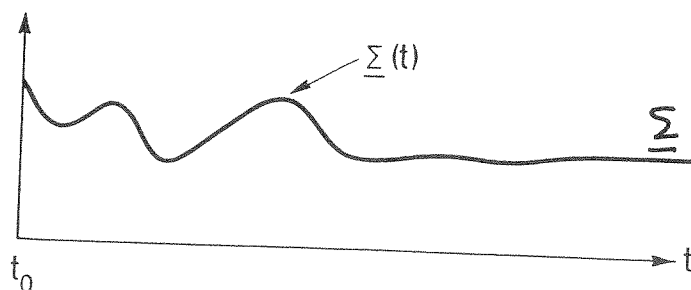
26

- If $[\underline{A}, \underline{L}]$ is controllable, then $\underline{\Sigma}$ is positive definite.

still solution $\neq$

$$\dot{\underline{\Sigma}}(t) = \underline{A} \underline{\Sigma}(t) + \underline{\Sigma}(t) \underline{A}' + \underline{L} \underline{\Xi} \underline{L}'$$

$$\underline{\Sigma}'(t_0) = \underline{\Sigma}_0$$



steady-state covariance matrix

27

FURTHER FACTS

- If initial state $\underline{x}(t_0)$ is gaussian.
- If white noise $\underline{\xi}(t)$ is gaussian.
- Then, $\underline{x}(t)$ is gaussian for all $t > 0$.

28

NUMERICAL EXAMPLE

- State Dynamics

$$\dot{\underline{x}}(t) = -\underline{x}(t) + \underline{\xi}(t) \quad (46)$$

$$\left. \begin{aligned} E\{x(0)\} &= 2 \\ \text{cov}[x(0); x(0)] &= 4 \end{aligned} \right\} \quad (47)$$

$$\left. \begin{aligned} E\{\xi(t)\} &= 0 \\ \text{cov}[\xi(t); \xi(\tau)] &= 6\delta(t-\tau) \end{aligned} \right\} \quad (48)$$

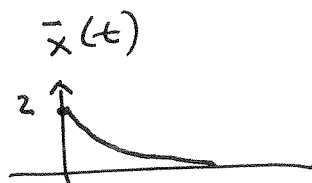
29

- Evolution of Mean

$$\bar{x}(t) = E\{x(t)\} \quad (49)$$

$$\dot{\bar{x}}(t) = -\bar{x}(t); \quad \bar{x}(0) = 2 \quad (50)$$

$$\rightarrow \boxed{\bar{x}(t) = 2e^{-t}} \quad (51)$$



30

- Evolution of Variance

$$\Sigma(t) = E\left\{\left(x(t) - \bar{x}(t)\right)^2\right\} \quad (52)$$

$$\dot{\Sigma}(t) = -2\Sigma(t) + 6; \quad \Sigma(0) = 4 \quad (53)$$

$$\rightarrow \boxed{\Sigma(t) = e^{-2t} + 3} \quad (54)$$

