

INPUT DESIGN ^{1.} (Ljung: Chap. 13). 5/13/02.

Input-output models (Scalar)

$$y(t) = G(q)u(t) + H(q)E(t) \quad \begin{array}{l} E(E(t)) = 0 \\ \text{Var}(E(t)) = \lambda \end{array}$$

$$G(q) = \sum_{k=1}^{\infty} g(k)q^{-k}$$

$$H(q) = 1 + \sum_{k=1}^{\infty} h(k)q^{-k}$$

G and H
stable filters.

Stable $\sum_{k=1}^{\infty} |g(k)| < \infty$

Family of Filters.

$$G_{\alpha}(q) = \sum_{k=1}^{\infty} g_{\alpha}(k)q^{-k} \quad \alpha \in A.$$

Uniform Stability

$$|g_{\alpha}(k)| \leq g(k) \quad \forall \alpha \in A$$

$$\sum_{k=1}^{\infty} g(k) < \infty.$$

Quasi-stationary Signals

Def: $(s(t))_{t=1}^{\infty}$ quasi-stationary

$$(i) \quad E s(t) = m_s(t) \quad |m_s(t)| \leq C \quad \forall t$$

2.

$$E(s(t)s(\tau)) = R_s(t, \tau) \quad |R_s(t, \tau)| \leq C$$

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N R_s(t, t-\tau) = R_s(\tau) \quad \forall \tau.$$

Expectation w.r.t measure corresponding to the stochastic part of signal.

Spectrum:

$$\Phi_s(\omega) = \sum_{\tau=-\infty}^{\infty} R_s(\tau) e^{-i\tau\omega}$$

Cross Spectrum

$$R_{sw}(\tau) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N R_{sw}(t, t-\tau)$$

$$R_{sw}(t, t-\tau) = E(s(t)w(t-\tau))$$

$$\Phi_{sw}(\omega) = \sum_{\tau=-\infty}^{\infty} R_{sw}(\tau) e^{-i\tau\omega}$$

$$R_s(0) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \Phi_s(\omega) d\omega$$

3.

Transformation of Spectra by Linear Systems

$$\Phi_y(\omega) = |G(e^{i\omega})|^2 \Phi_u(\omega) + \lambda |H(e^{i\omega})|^2$$

$$\Phi_{y_u}(\omega) = G(e^{i\omega}) \Phi_u(\omega).$$

Ergodicity (Ljung: P. 43).

Th: Let $(s(t))_{t=1}^{\infty}$ be quasi-stationary signal,

$E(s(t)) = m(t)$. Assume

$$s(t) - m(t) = v(t) = \sum_{k=0}^{\infty} h_t(k) e(t-k) = H_t(q) E(t),$$

$E(t)$: zero-mean, independent $\text{Var}(E(t)) = 1$,

and bounded fourth moments and $(H_t(q) | t=1, 2, \dots)$

uniformly stable family of filters. Then with

prob. 1 as $N \rightarrow \infty$

$$(i) \quad \frac{1}{N} \sum_{t=1}^N s(t) s(t-\tau) \longrightarrow R_s(\tau), \text{ where}$$

$$R_s(\tau) := \overline{E} s(t) s(t-\tau) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N E(s(t))$$

$$(ii) \quad \frac{1}{N} \sum_{t=1}^N [s(t) m(t-\tau) - E s(t) m(t-\tau)] \longrightarrow 0$$

4.

$$\frac{1}{N} \sum_{t=1}^N [s(t)v(t-\tau) - E(s(t)v(t-\tau))] \rightarrow 0.$$

Remark: Spectrum of an observed single realization of $(s(t))$ computed as a deterministic signal, coincides with prob. 1 with that of the process $(s(t))$ defined in terms of an ensemble average (Note: second order property).

One-step ahead Prediction:

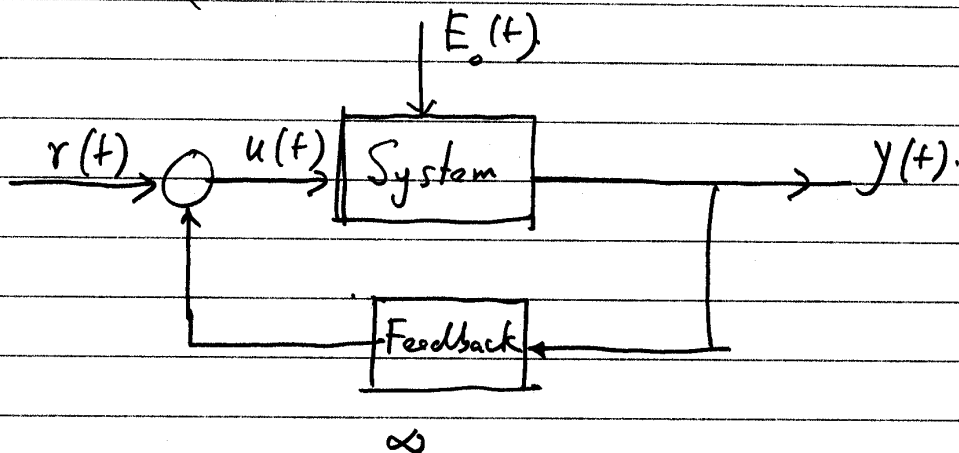
$H(q)$ invertible $\left(\frac{1}{H(z)} \text{ analytic in } |z| \geq 1\right)$
 $(\Rightarrow \bar{H}'(z) \text{ stable})$

$$\hat{Y}(t|t-1) = \bar{H}'(q) G(q) u(t) + [1 - \bar{H}'(q)] Y(t).$$

5.

Data Set-

$$\mathcal{Z}^N = (u(1), y(1), \dots, u(N), y(N))$$



Data set $\mathcal{Z}$

$$Y(t) = \sum_{k=1}^{\infty} d_t^{(1)}(k) r(t-k) + \sum_{k=0}^{\infty} d_t^{(2)}(k) E_o(t-k)$$

$$U(t) = \sum_{k=0}^{\infty} d_t^{(3)}(k) r(t-k) + \sum_{k=0}^{\infty} d_t^{(4)}(k) E_o(t-k)$$

- 1) $r(t)$: bounded, deterministic,
- 2) $E_o(t)$: zero-mean, independent : odd for the order moments.
- 3) $d_t^{(i)}(k)$: uniformly stable.
- 4) $(Y(t), U(t))$: quasi-stationary

6.

True system:

$$S: Y(t) = G_o(q) u(t) + H_o(q) E_o(t).$$

satisfies requirements of data set Z^∞ .

$$\mathcal{M}: (G(q, \theta), H(q, \theta) \mid \theta \in \mathcal{D}_M).$$

$S \in \mathcal{M}$: hypothesis

Inputs generated by:

$$u(t) = -F(q)Y(t) + r(t).$$

s.t. conditions on Z^∞ - hold.

Def: A quasi-stationary data set Z^∞

is informative enough with. r. t. model set $\mathcal{M}^*$

defined by the model structure if for any

two models $W_1(q)$ and $W_2(q)$

$$\overline{E} [(W_1(q) - W_2(q)) Z(t)] = 0 \Rightarrow W_1(e^{i\omega}) \Big|_{\substack{\omega \in \omega_1 \\ \omega \in \omega_2}} = W_2(e^{i\omega}).$$

7.

Proposition: $\sum_{t=0}^{\infty}$ informative if spectrum matrix of $z(t) = (y(t), u(t))$ strictly pos. def.

$\forall \omega \in \mathbb{R}$

Informative Experiments.

Let $\theta_1, \theta_2 \in \mathbb{D}_M$

$$\varepsilon_i(t) = \varepsilon(t, \theta_i) \quad i=1,2.$$

$$\Delta \varepsilon(t) = \varepsilon_1(t) - \varepsilon_2(t) = \frac{1}{H_1(q)} \left[\Delta G(q) u(t) + \Delta H(q) \varepsilon_2(t) \right]$$

$$\varepsilon_2(t) = \frac{1}{H_2(q)} \left[(G_0(q) - G_2(q)) u(t) + H_0(q) \varepsilon_0(t) \right]$$

$$\overline{E} \left(\Delta \varepsilon(t) \right)^2 = \frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{1}{|H_1(e^{i\omega})|^2} \left[\left| \Delta G(e^{i\omega}) + \frac{G_0(\cdot) - G_2(\cdot)}{H_2(\cdot)} + \Delta H(\cdot) \right|^2 \right] d\omega$$

$$+ \left| \Delta H(\cdot) \right|^2 \left| \frac{H_0(\cdot)}{H_2(\cdot)} \right|^2 \lambda_0 \right] d\omega$$

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$$\text{Now } |H_0(e^{i\omega})|^2 > 0 \quad (\text{Invertibility}) \\ \forall \omega$$

If data not informative w.r. to M^*

$$\bar{E} \left(\Delta \varepsilon(t) \right)^2 = 0 \quad (1)$$

$$\Rightarrow \Delta H(e^{i\omega}) \equiv 0 \quad (2)$$

$$\Rightarrow |\Delta G(e^{i\omega})|^2 \Phi_u(\omega) = 0 \quad (3)$$

If (3) $\Rightarrow \Delta G \equiv 0$, then it follows from (1)

that the two models are equal.

$\Rightarrow$ data suff. inf. w.r. to M^* .

Def: A quasi-stationary $(u(t))_{t=1}^{\infty}$ with spectrum $\Phi_u(\omega)$ is said to be persistently

exciting of order n if $\forall$ filters

$$M_n(q) = m_1 \bar{q}^1 + \dots + m_n \bar{q}^n$$

9.

$$|M_n(e^{i\omega})|^2 \Phi_n(\omega) \equiv 0 \Rightarrow M_n(e^{i\omega}) \equiv 0$$

$(\Rightarrow \Phi_n(\omega) \neq 0$ on at least n -points $-\pi < \pi < \pi$

Equivalent characterization: $R_n(\tau)$.

$$\bar{R}_n = \begin{bmatrix} R_n(0) & \dots & R_n(n-1) \\ R_n(1) & R_n(0) & \dots & R_n(n-2) \\ \vdots & \vdots & \ddots & \vdots \\ R_n(n-1) & R_n(n-2) & \dots & R_n(0) \end{bmatrix} \text{ non-singular.}$$

Th: Q For

$$G(q, \theta) = \frac{B(q, \theta)}{F(q, \theta)} = \frac{q^{-n_b} (b_1 + b_2 q^{-1} + \dots + b_{n_b} q^{-n_b+1})}{1 + f_1 q^{-1} + \dots + f_{n_f} q^{-n_f}}$$

Open loop expt. with input persistently
exciting of order $n_b + n_f$ suff. inf.

10.

Closed-loop Identification.

$$Y(t) = G_o(q)u(t) + \underbrace{H_o(q)}_{v(t)}E(t)$$

$$u(t) = r(t) - F_y(q)y(t). \quad \text{Closed-loop system stable!}$$

Closed-loop eqns.

$$Y(t) = G_o(q)S_o(q)r(t) + S_o(q)v(t).$$

$$u(t) = S_o(q)r(t) - F_y(q)S_o(q)v(t).$$

$$S_o(q) = \frac{1}{1 + F_y(q)G_o(q)}.$$

Input Spectrum

$$\begin{aligned} \Phi_u &= |S_o|^2 \Phi_r + |F_y|^2 |S_o|^2 \Phi_v \\ &= \Phi_u^r + \Phi_u^e \end{aligned}$$

Convergence Th. holds: require true system contained in Model set.

11.

Input persistently exciting $\not\Rightarrow$ necessarily closed-loop system. inf.

Ex:

$$Y(t) + aY(t-1) = bU(t-1) + E(t)$$

$$u(t) = -fY(t) \quad (\text{delay in feedback}).$$

$$\Rightarrow Y(t) + (a+bf)Y(t-1) = E(t) \quad (\text{Closed loop}).$$

$\Rightarrow$ all models with

$$\begin{aligned} \hat{a} &= a + rf \\ \hat{b} &= b - r \end{aligned} \quad r \in \mathbb{R}$$

gives same input-output reln. as the system (a, b) with feedback.

$\Rightarrow$ Cannot distinguish between models.

Soln: Persist model

$$Y(t) + aY(t-1) = u(t-1) + E(t)$$

Suff. inf. with respect to a .

12.

Informative Closed-loop Experiments

Defn: on P.6.

$$\overline{E} \left[\Delta W_y(q) y(t) + \Delta W_u(q) u(t) \right]^2 = 0.$$

For the regulator.

$$\begin{aligned} 0 &= \overline{E} \left| \begin{bmatrix} \Delta W_y & \Delta W_u \end{bmatrix} \begin{bmatrix} G_o & S_o \\ 1 & -F_y S_o \end{bmatrix} \begin{bmatrix} S_o r \\ v \end{bmatrix} \right|^2. \\ &= \overline{E} \left| \begin{bmatrix} \tilde{W}_y & \tilde{W}_u \end{bmatrix} \begin{bmatrix} S_o r \\ v \end{bmatrix} \right|^2. \\ &= \overline{E} \left| \tilde{W}_y S_o r \right|^2 + \overline{E} \left| \tilde{W}_u v \right|^2. \end{aligned}$$

v : persistently exciting

If $S_o r$ is also persistently exciting

$$\Rightarrow [\tilde{W}_y, \tilde{W}_u] \equiv 0 \Rightarrow \Delta W_y = 0 \text{ and } \Delta W_u = 0$$