

System Identification

6.435

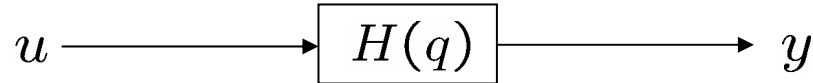
SET 1

- Review of Linear Systems
- Review of Stochastic Processes
- Defining a General Framework

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Review

LTI discrete-time systems



$$y(t) = \sum_{k=0}^{\infty} h(t-k)u(k)$$

$$H = \text{transfer function} = \sum_{t=0}^{\infty} h(t)q^{-t}$$

Note ($q \simeq z \simeq \lambda$ different notations)

Stability

$$\sum_{t=0}^{\infty} |h(t)| < \infty$$

\Leftrightarrow (real rational $H(q)$) poles of H are outside the disc

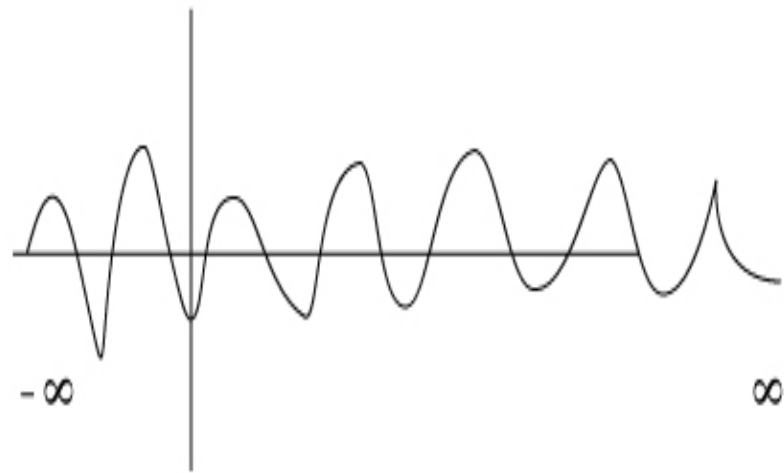
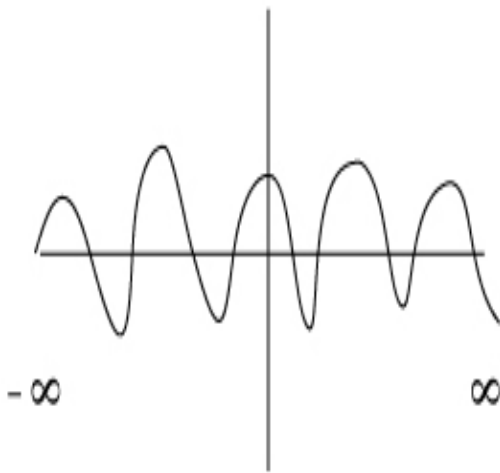
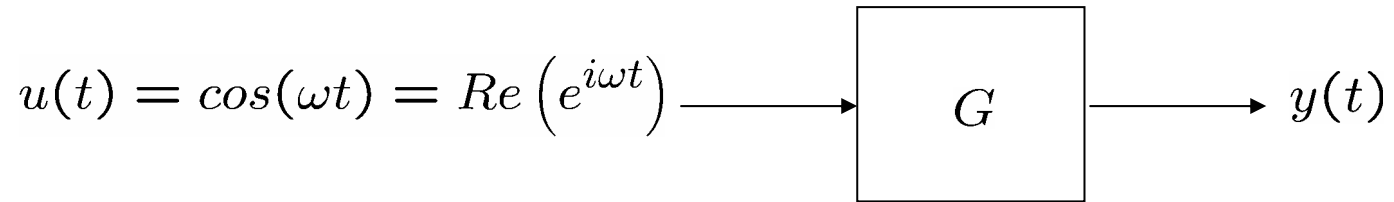
Strict causality

$h(0) = 0$ system has a delay.

Strict Stability

$$\sum_{t=0}^{\infty} t |h(t)| < \infty$$

Frequency Response



$$\begin{aligned}
y(t) &= \operatorname{Re} \left[\sum_{k=0}^{\infty} g(k) e^{i\omega(t-k)} \right] \\
&= \operatorname{Re} \left[e^{i\omega t} \left(\sum_{k=0}^{\infty} g(k) e^{-i\omega k} \right) \right] \\
&= \operatorname{Re} \left[e^{i\omega t} G(e^{i\omega}) \right] \\
&= |G(e^{i\omega})| \cos(\omega t + \operatorname{avg} G(e^{i\omega})).
\end{aligned}$$

Suppose $u(t) = 0 \quad \forall \quad t < 0$, then

$$y(t) = |G(e^{i\omega})| \cos(\omega t + \text{arg}G(e^{i\omega})) - \underbrace{\text{Re} \left[e^{i\omega t} \sum_{k=t}^{\infty} g(k) e^{-i\omega k} \right]}_{\leq \sum_{k=t}^{\infty} |g(k)| \rightarrow 0}$$

for stable systems.

Periodograms

Given: $\{u(t), t = 1, 2, \dots, N\}$, the Fourier transform is given by

$$U_N(\omega) = \frac{1}{\sqrt{N}} \sum_{t=1}^N u(t) e^{-i\omega t}$$

Discrete – Fourier Transform (DFT)

$$\left\{ U_N \left(\frac{2\pi}{N} k \right), k = 1, \dots, N \right\}.$$

Inverse DFT

$$u(t) = \frac{1}{\sqrt{N}} \sum_{k=1-\frac{N}{2}}^{\frac{N}{2}} U_N \left(\frac{2\pi}{N}k \right) e^{i \left(\frac{2\pi}{N}t \right) - k}$$

Periodogram

$$= |U_N(\omega)|^2$$

Nice Property: Parsaval's equality

$$\sum_{k=1}^N \left| U_N \left(\frac{2\pi}{N}k \right) \right|^2 = \sum_{t=1}^N |u(t)|^2$$

$$\left| U_N \left(\frac{2\pi}{N} k \right) \right|^2 = \text{energy contained in each frequency component.}$$

Properties:

-periodicity $U_N(\omega + 2\pi) = U_N(\omega)$

-If $u(t)$ is real $U_N(-\omega) = \overline{U_N(\omega)}$

Example: $u(t) = A \cos(\omega_0 t)$ $\omega_0 = \frac{2\pi}{N_0}$

(consider $N = sN_0$)

$$\begin{aligned}
U_N(\omega) &= \frac{1}{\sqrt{N}} \sum_{t=1}^N A \cos\left(\frac{2\pi}{N_0}t\right) e^{-i\omega t} \\
&= \frac{1}{\sqrt{N}} \sum_{t=1}^N \frac{A}{2} \left[e^{i\omega_0 t} + e^{-i\omega_0 t} \right] e^{-i\omega t} \\
&= \frac{1}{\sqrt{N}} \sum_{t=1}^N \frac{A}{2} \left[e^{i(\omega_0 - \omega)t} + e^{-i(\omega_0 + \omega)t} \right] \\
&= \begin{cases} \frac{\sqrt{N}A}{2} & \text{if } \omega = \pm\omega_0 = \frac{2\pi}{N_0} \\ 0 & \text{if } \omega = \frac{2\pi}{N}k, k \neq s \\ ? & \omega - \text{other} \end{cases}
\end{aligned}$$

$$|U_N(\omega)|^2 = \begin{cases} \frac{A^2}{4}N & \text{if } \omega = \pm\omega_o = \frac{2\pi}{N_o} = \frac{2\pi s}{N} \\ 0 & \text{if } \omega = \frac{2\pi}{N}k = \frac{2\pi}{sN_o}k, k \neq s \\ ? & \text{other.} \end{cases}$$

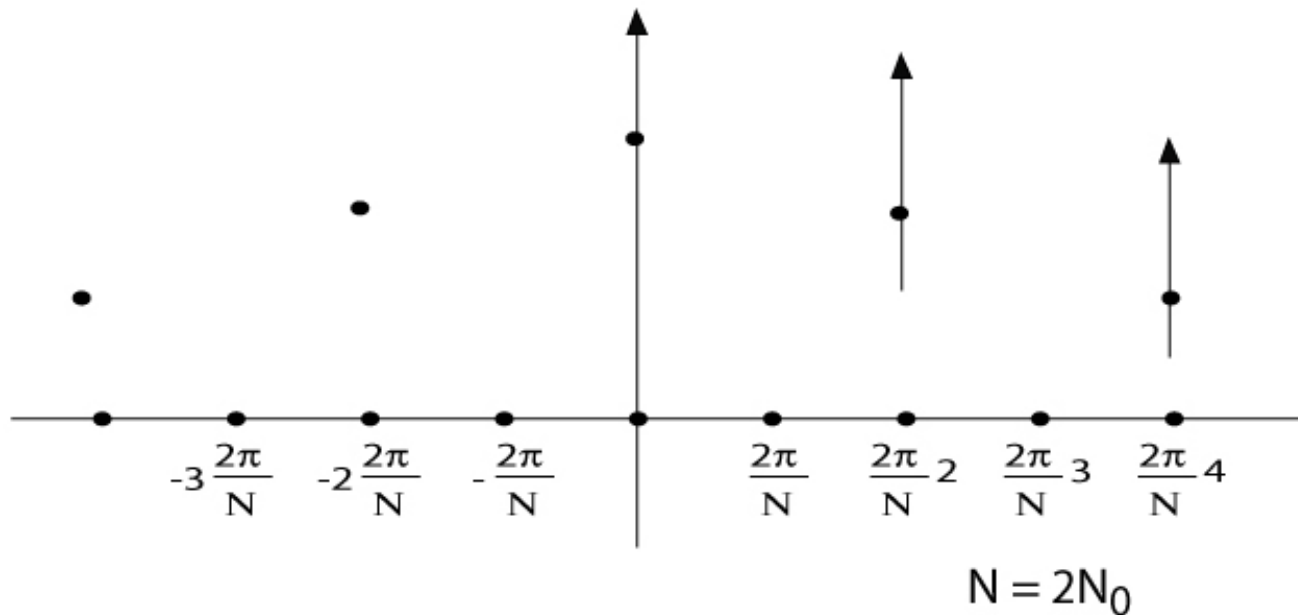
Example: $u(t) = (t + N_o) \quad N = sN_o$

$$u(t) = \frac{1}{\sqrt{N_o}} \sum_{r=-\frac{N_o}{2}+1}^{\frac{N_o}{2}} A_r \left(e^{i\frac{2\pi}{N_o}t} \right) r \quad 0 \leq t \leq N_o$$

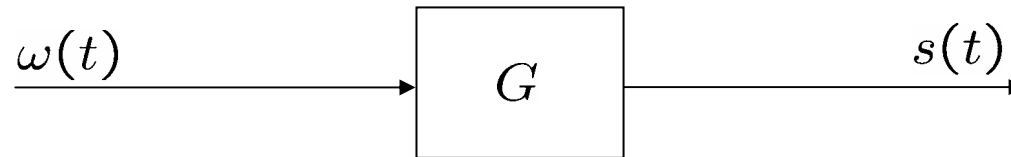
(Why?) $A_r = \frac{1}{\sqrt{N_o}} \sum_{t=1}^{N_o} u(t) \left(e^{-\frac{2\pi i}{N_o}r} \right) t$

This representation is valid over the whole interval $[1, N]$ since u is periodic over N .

$$|U_N(\omega)|^2 = \begin{cases} s^2 |A_r|^2 & \text{if } \omega = \frac{2\pi}{N_0} r, r \neq \pm 0, \pm 1, \dots \pm \frac{N_0}{2} \\ 0 & \text{if } \omega = \frac{2\pi}{N} k, k \neq r \cdot s \\ ? & \omega \text{ other.} \end{cases}$$



Passing through a filter



$$|\omega(t)| \leq C_W \quad \forall \quad t \in \mathbb{R} \quad C_G = \sum_{k=1}^N k |g(k)| < \infty$$

Claim:

$$S_N(\omega) = G(e^{i\omega}) W_N(\omega) + R_N(\omega)$$

$$|R_N(\omega)| \leq 2 \frac{C_W C_G}{\sqrt{N}}$$

Proof:

$$\begin{aligned} S_N(\omega) &= \frac{1}{\sqrt{N}} \sum_{t=1}^N s(t) e^{-i\omega t} = \frac{1}{\sqrt{N}} \sum_{t=1}^N \sum_{k=1}^{\infty} g(k) \omega(t-k) e^{-i\omega t} \\ &= \frac{1}{\sqrt{N}} \sum_{k=1}^{\infty} g(k) \sum_{t=1}^N \omega(t-k) e^{-i\omega t} \\ &= \frac{1}{\sqrt{N}} \sum_{k=1}^{\infty} g(k) e^{-i\omega k} \sum_{\tau=1-k}^{\tau=N-k} \omega(\tau) e^{-i\omega \tau} \end{aligned}$$

Note that:

$$\left| W_N(\omega) - \frac{1}{\sqrt{N}} \sum_{\tau=1-k}^{N-k} \omega(\tau) e^{-i\omega \tau} \right|$$

$$\leq \left| \frac{1}{\sqrt{N}} \sum_{\tau=1-k}^0 \omega(\tau) e^{-i\omega\tau} \right| + \left| \frac{1}{\sqrt{N}} \sum_{\tau=N-k+1}^N \omega(\tau) e^{-i\omega\tau} \right|$$

$$\leq 2kC_W \frac{1}{\sqrt{N}}$$

Hence:

$$\left| S_N(\omega) - G(e^{i\omega}) W_N(\omega) \right|$$

$$= \left| \sum_{k=1}^{\infty} g(k) e^{-i\omega k} \left[\frac{1}{\sqrt{N}} \sum_{\tau=1-k}^{N-k} \omega(\tau) e^{-i\omega\tau} - W_N(\omega) \right] \right|$$

$$\leq \frac{2C_W}{\sqrt{N}} \sum_{k=1}^{\infty} |g(k)| \cdot k = 2 \frac{C_W C_G}{\sqrt{N}}$$

Stochastic Processes

Definition:

A stochastic process is a sequence of random variables with a joint pdf.

Definition:

$$m_x(t) = Ex(t) = \text{mean}$$

$$R_x(t) = Ex(t_1)x^T(t_2) = \text{Correlation / Covariance}$$

$$C_x(t_1, t_2) = E(x(t_1) - m_x(t_1))(x(t_2) - m_x(t_2))^T \\ = \text{Covariance}$$

$$\text{Var}(x(t)) = R_x(t, t) - E^2(x(t))$$

Traditional definitions:

$x(t)$ is Wide-sense stationary (WSS) if

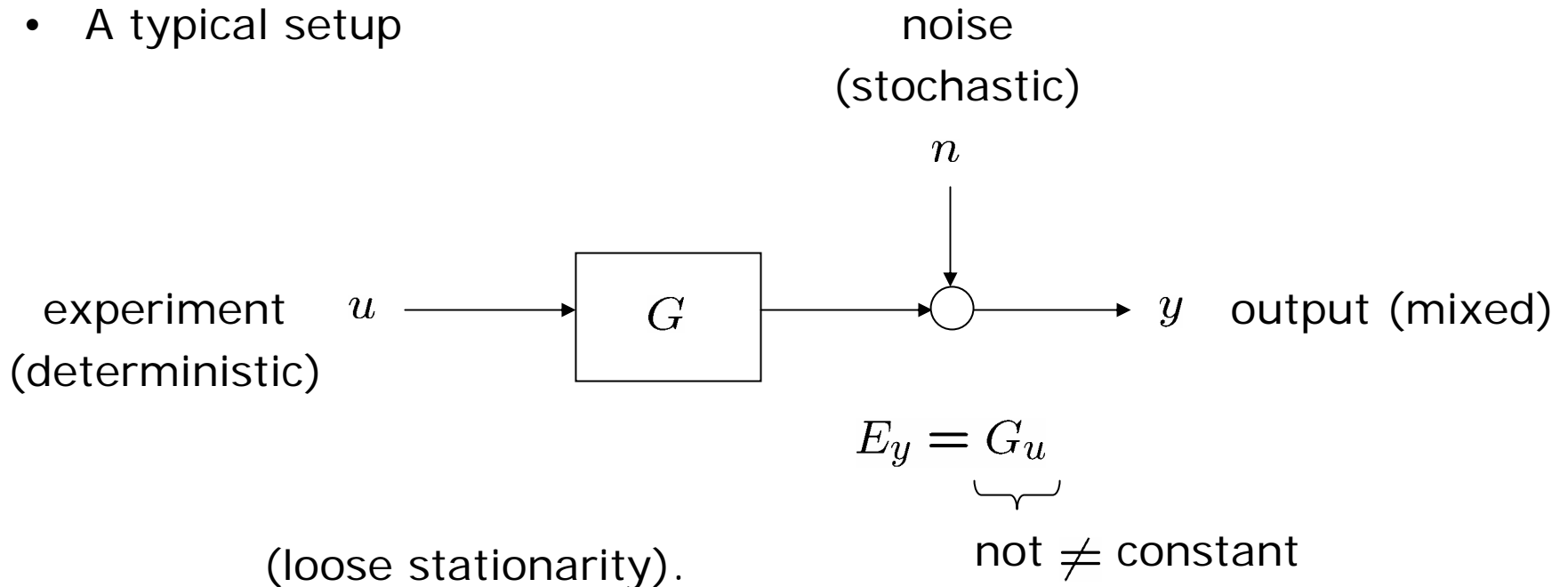
$$m_x(t) = m_x = \text{constant} \quad \forall \quad t$$

$$R_x(t_1, t_2) = R_x(t_1 - t_2)$$

“This may be a limiting definition.” We will discuss shortly.

Common Framework for Deterministic and Stochastic Signals

- A typical setup



- Consider signals with the following assumptions:

$$1) \quad E s(t) = m_s(t) \quad |m_s(t)| \leq C \quad \forall \quad t$$

$$2) \quad E s(t)s(r) = R_s(t, r) \quad |R_s(t, r)| \leq C$$

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

$s(t)$ is called quasi-stationary

- If $s(t)$ is a stationary process, then it satisfies 1, 2 trivially.
- If $s(t)$ is a deterministic signal, then

$$1) \quad |s(t)| \leq C$$

$$2) \quad \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N s(t)s(t - \tau) = R_s(\tau)$$

- Example: Suppose $s(t)$ has finite energy

$$\left| \frac{1}{N} \sum_{t=1}^N u(t)u(t-\tau) \right| \leq \frac{1}{N} \|u\|_2^2 \rightarrow 0 \quad N \rightarrow \infty$$

- In general: $s(t) = \underbrace{x(t)}_{\text{stochastic}} + \underbrace{u(t)}_{\text{deterministic}}$

- Notation: $\bar{E}(\cdot) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N (\cdot)$

quasi-stationary $\cong \bar{E}s(t) = m_s = \text{constant}$

$$\bar{E}s(t)s(\tau) = R_s(t - \tau)$$

- $s(t) = x(t) + u(t); \quad \bar{E}s(t) = m_x + m_u$
 $\bar{E}s(t)s(t - \tau) = R_x(\tau) + R_u(\tau) + 2m_x m_u$

Power Spectrum

Let $x(t)$ be a quasi-stationary process. The power spectrum $\Phi_x(e^{i\omega})$ is defined as

$$\begin{aligned}\Phi_x(e^{i\omega}) &= \sum_{\tau=-\infty}^{\infty} R_x(\tau)e^{-i\omega\tau} \\ &= \text{Fourier transform of } R_x\end{aligned}$$

- Example: for $x(t) = A\cos(\omega_0 t)$

$$\begin{aligned}R_x(\tau) &= \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N A^2 \cos\omega_0 t \cdot \cos\omega_0(t - \tau) \\ &= \frac{A^2}{2} \cos(\omega_0 \tau)\end{aligned}$$

$$\Phi_x(e^{i\omega}) = \frac{A^2}{4} (\delta(\omega - \omega_0) + \delta(\omega + \omega_0))$$

Recall

$$|X_N(\omega)|^2 = \begin{cases} \frac{A^2}{4}N & \text{if } \omega = \pm\omega_0 \\ (0, ?) & \text{other} \end{cases}$$

$$\lim_{N \rightarrow \infty} |X_N(\omega)|^2 \xrightarrow{?} \Phi_x(e^{i\omega})$$

- Result ; for any $s(t)$, quasi-stationary

$$E \int_{-\pi}^{\pi} |S_N(\omega)|^2 \Psi(\omega) d\omega \xrightarrow[N \rightarrow \infty]{} \int_{-\pi}^{\pi} \Phi_s(e^{i\omega}) \Psi(\omega) d\omega$$

Convergence as a distribution

Note: $|S_N(\omega)|^2$ an erratic function

$\Phi_s(e^{i\omega})$ well behaved function

Cross Spectrum

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

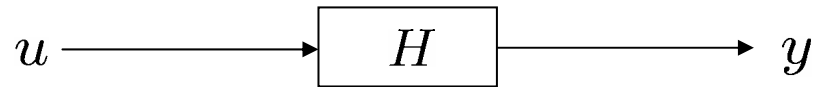
Spectrum of mixed signals

$$s(t) = x(t) + u(t) \quad \{\text{zero means}\}$$

$$R_s = R_x + R_u$$

$$\Phi_s = \Phi_x + \Phi_u$$

Spectrum of Filtered Signals



$$\Phi_y(e^{i\omega}) = H(e^{i\omega}) \Phi_u(e^{i\omega}) H^*(e^{i\omega})$$
$$\stackrel{\text{SISO}}{=} \Phi_u(e^{i\omega}) |H(e^{i\omega})|^2$$

If $u = WN$ signal $\phi_u = \lambda^2 I$

$$\Phi_y(e^{i\omega}) = \lambda^2 H H^*(e^{i\omega})$$
$$\stackrel{\text{SISO}}{=} \lambda^2 |H(e^{i\omega})|^2$$

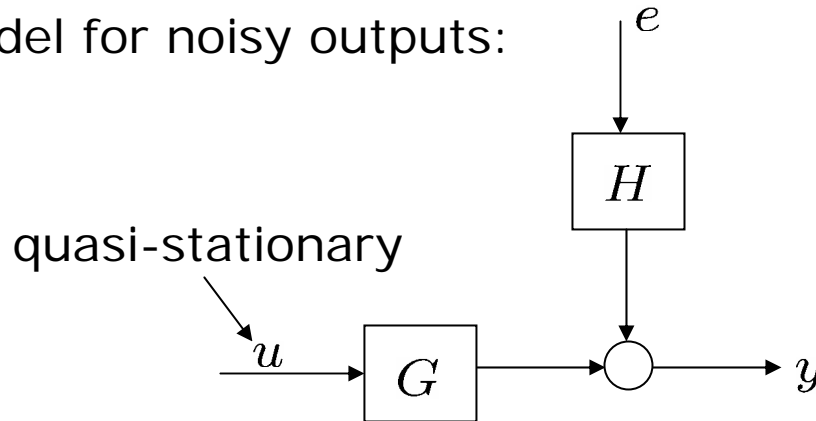
Generation of a process with a given Covariance:

given $R_x(\tau)$, then x is the output of a filter with a WN signal as an input. The Filter is the spectral factor of $\Phi_x(e^{i\omega})$;

$$\Phi_x(e^{i\omega}) = \underbrace{H(e^{i\omega})}_{\text{stable minimum phase}} \cdot H(e^{i\omega})^*$$

Important relations

A model for noisy outputs:



$$y = Gu + He$$

$$E(e(t)e^T(t-\tau)) = \underbrace{\Lambda^2 \delta(\tau)}_{\text{constant.}}$$

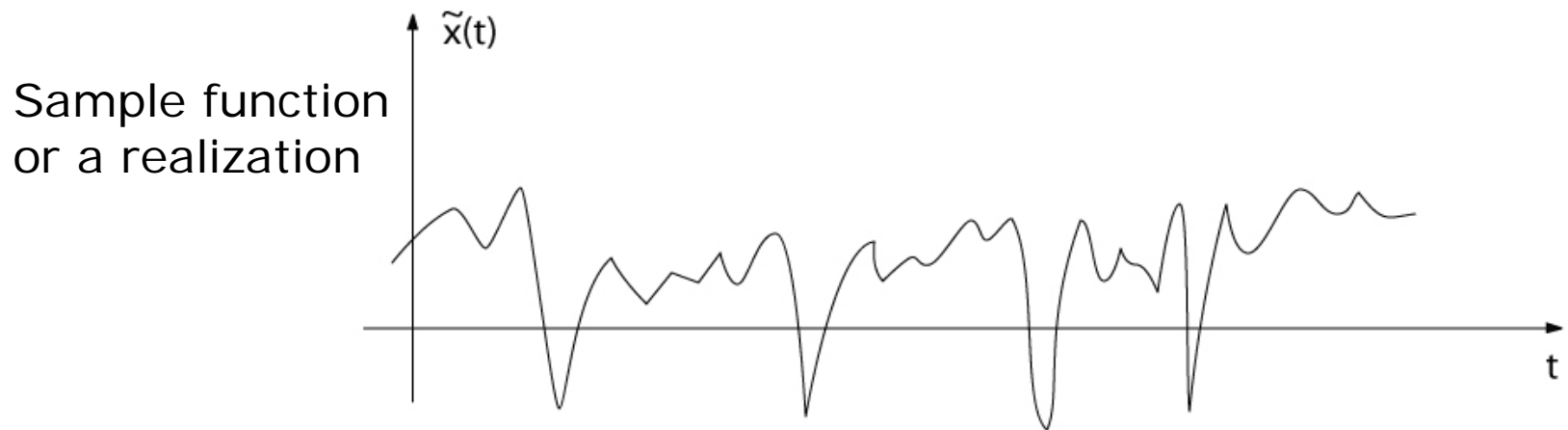
$$\Phi_y(e^{i\omega}) = G\Phi_u(e^{i\omega})G^T(e^{i\omega})H\Lambda^2H^*(e^{i\omega})$$

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

- Very Important relations in system ID.
- Correlation methods are central in identifying an unknown plant.
- Proofs: Messy; Straight forward.

Ergodicity

- $x(t)$ is a stochastic process



- Sample mean = $\bar{E}\tilde{x}(t) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N \tilde{x}(t)$

- Sample Covariance = $\bar{E}\tilde{x}(t)\tilde{x}(t - \tau) = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{t=1}^N \tilde{x}(t)\tilde{x}(t - \tau)$

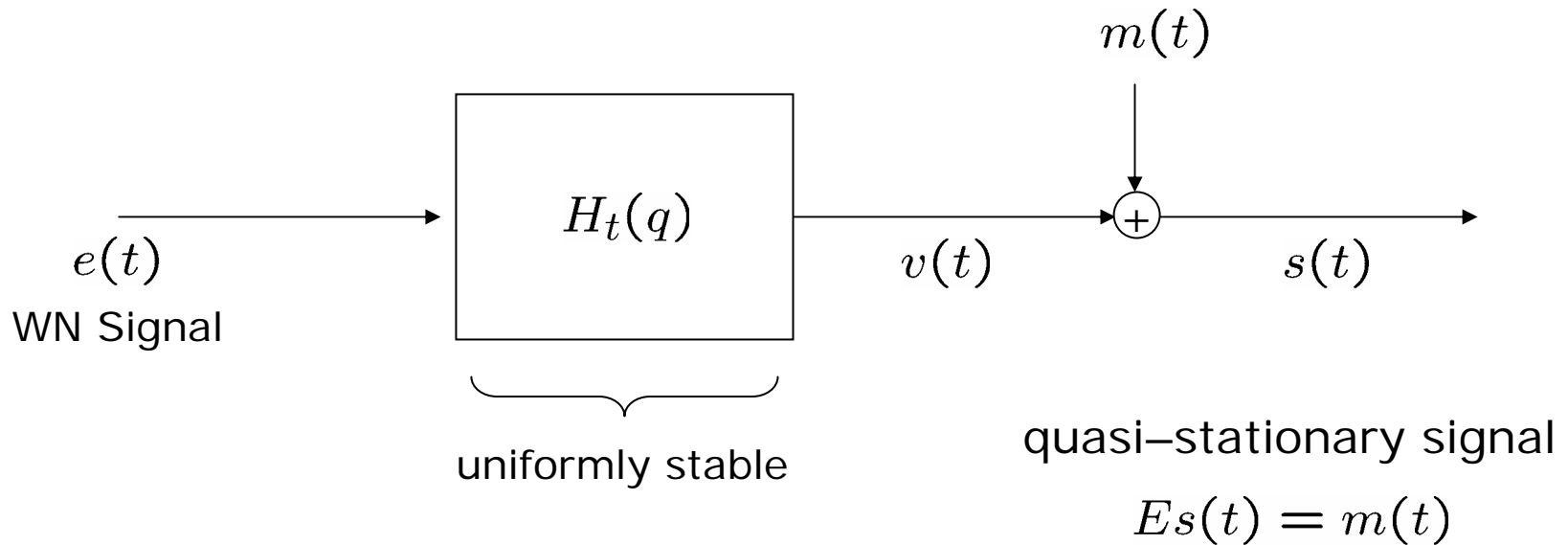
- A process is 2nd-order ergodic if

mean $\hat{=} Ex(t) =$ the sample mean of any realization.

covariance $\hat{=} Ex(t)x(t - \tau) =$ the sample covariance of any realization.

- Sample averages \simeq Ensemble averages

A general ergodic process



$$\bar{E}s(t)s(t - \tau) = R_s(\tau) \quad \text{w.p.1}$$

$$\frac{1}{N} \sum_{t=1}^N [s(t)m(t - \tau) - Es(t)m(t - \tau)] \rightarrow 0 \quad \text{w.p.1}$$

$$\frac{1}{N} \sum_{t=1}^N [s(t)v(t - \tau) - Es(t)v(t - \tau)] \rightarrow 0 \quad \text{w.p.1}$$

Remark:

Most of our computations will depend on a given realization of a quasi-stationary process. Ergodicity will allow us to make statements about repeated experiments.