

### (5) Scaling property

if  $g(t) \leftrightarrow G(\omega)$  then

for any const  $a$ ,  $g(at) \leftrightarrow \frac{1}{|a|} G\left(\frac{\omega}{a}\right)$

Proof:

for +ve const  $a$ ,

$$f(g(at)) = \int_{-\infty}^{\infty} g(at) e^{-j\omega t} dt$$

$$= \frac{1}{a} \int_{-\infty}^{\infty} g(x) e^{-j\omega/a x} dx \quad x = at$$

$$= \frac{1}{a} G\left(\frac{\omega}{a}\right)$$

if  $a < 0$ ,

$$g(at) \leftrightarrow -\frac{1}{a} G\left(\frac{\omega}{a}\right)$$

Convolution

The convolution of two func  $g(t)$  &  $w(t)$ , denoted by  $g(t) * w(t)$  is,

$$g(t) * w(t) = \int_{-\infty}^{\infty} g(t) w(t-\tau) dt$$

$$\text{if } g_1(t) \leftrightarrow G_1(\omega) \text{ & } g_2(t) \leftrightarrow G_2(\omega)$$

$$\text{then, } g_1(t) * g_2(t) \xrightarrow{\text{Time}} G_1(\omega) G_2(\omega) \quad [\text{Freq}]$$

by defn,

$$\begin{aligned} F[g_1(t) * g_2(t)] &= \int_{-\infty}^{\infty} e^{-j\omega t} \left[ \int_{-\infty}^{\infty} g_1(\tau) g_2(t-\tau) d\tau \right] dt \\ &= \int_{-\infty}^{\infty} g_1(\tau) \left[ \int_{-\infty}^{\infty} e^{-j\omega t} g_2(t-\tau) dt \right] d\tau \\ &= \int_{-\infty}^{\infty} g_1(\tau) e^{-j\omega \tau} G_2(\omega) d\tau \\ &= G_2(\omega) \int_{-\infty}^{\infty} g_1(\tau) e^{-j\omega \tau} d\tau = G_2(\omega) G_1(\omega) \end{aligned}$$

Thm:  $g(t) \leftrightarrow G(\omega)$

$$\int_{-\infty}^t g(\tau) d\tau \leftrightarrow \frac{G(\omega)}{j\omega} + \pi G(0) \delta(\omega)$$

Proof

⑥ Time conv.

$$g(t) * u(t) = \int_{-\infty}^{\infty} g(\tau) u(t-\tau) d\tau = \int_{-\infty}^t g(\tau) dt$$

$$g(t) * (u(t)) \leftrightarrow G(\omega) U(\omega)$$

$$= G(\omega) \left[ \frac{1}{j\omega} + \pi \delta(\omega) \right]$$

$$= \frac{G(\omega)}{j\omega} + \pi G(0) \delta(\omega)$$

⑦ Time diff & integration

$$g(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} G(\omega) e^{j\omega t} d\omega$$

$$\frac{dg}{dt} = \frac{1}{2\pi} \int_{-\infty}^{\infty} j\omega G(\omega) e^{j\omega t} d\omega$$

$$\frac{dg}{dt} \leftrightarrow j\omega G(\omega)$$

$$\frac{d^n g}{dt^n} \leftrightarrow j\omega^n G(\omega)$$

Properties

- ① Commutative
- ② Associative
- ③ Distributive

if  $\int g(t) dt \rightarrow$  real  
 $G_r(w) = G_r^\infty(w)$

$$\int_{-\infty}^{\infty} g(t) e^{-j\omega t} dt = G_r(w)$$

$$\frac{1}{2\pi} \int_{-\infty}^{\infty} g(t) e^{-j\omega t} dt = g(w)$$

### Properties

#### ① Linearity

$$x_1(t) \rightarrow X_1(j\omega)$$

$$x_2(t) \rightarrow X_2(j\omega)$$

$$\alpha x_1(t) + \beta x_2(t) \rightarrow \alpha X_1(j\omega) + \beta X_2(j\omega)$$

#### Proof

$$x(t) \leftrightarrow X(j\omega)$$

$$X(j\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$$

$$= \int_{-\infty}^{\infty} [\alpha x_1(t) + \beta x_2(t)] e^{-j\omega t} dt$$

$$= \alpha X_1(j\omega) + \beta X_2(j\omega)$$

#### ② Conjugation

$$x(t) \leftrightarrow X(j\omega)$$

$$x^*(t) \leftrightarrow X^*(j\omega)$$

(3) Area under  $x(t)$

$$\int x(t) dt$$

$$X(0) = \int_{-\infty}^{\infty} x(t) dt \quad \text{at } w=0$$

(4) Area under  $x(w)$

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \dots dt$$

$$x(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} x(w) dw \quad t=0 \quad (w) \rightarrow (0) x$$

$$(2\pi x(0))$$

(5) Time reversal

$$x(t) \Leftrightarrow x(iw)$$

$$x(-t) \Leftrightarrow x(-iw)$$

Proof  $X(w) = \int_{-\infty}^{\infty} x(t) e^{-jwt} dt$

$$= \int_{-\infty}^{\infty} x(-t) e^{jw(-t)} dt$$

$$= \int_{-\infty}^{\infty} x(t) e^{-jwt} dt$$

$$= X(-w) = \dots$$

## 6) Time scaling

$$x(\alpha t) = \frac{1}{|\alpha|} x\left(\frac{\omega}{\alpha}\right)$$

$$\text{C} x(\omega) = \int_{-\infty}^{\infty} x(\alpha t) e^{-j\omega t} dt$$

$$x(\omega) = \int_{-\infty}^{\infty} x(\gamma) e^{-j\omega t/a} \frac{d\gamma}{a}$$

$$= \frac{1}{a} \int x(\gamma) e^{-j\omega/a \gamma} d\gamma$$

$$x(\omega) = \frac{1}{a} x\left(\frac{j\omega}{a}\right)$$

$$x(\omega) = \frac{1}{a} x\left(\frac{j\omega}{a}\right)$$

## 7) Time shifting

$$x(t+t_0) \Leftrightarrow x(j\omega) e^{j\omega t_0}$$

proof:

$$\int x(t+t_0) e^{-j\omega t} dt$$

$$= \int x(t) e^{-j\omega(t-t_0)} dt$$

$$= e^{j\omega t_0} x(\omega)$$

## 8) Freq shifting

$$x(t) \Leftrightarrow x(j\omega)$$

$$e^{\pm j\omega_0 t} x(t) \Leftrightarrow x(\omega \pm \omega_0)$$

9) Conv in time

$$x_1(t) * x_2(t) \leftrightarrow X_1(\omega) * X_2(\omega)$$

Proof given earlier

10) Duality

$$X(t) \leftrightarrow 2\pi X(-\omega)$$

$$X(-\omega) = \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt$$

$$2\pi X(-\omega) = X(t)$$

FTs①  $\delta(t)$ 

$$\begin{aligned}
 x(\omega) &= \int_{-\infty}^{\infty} x(t) e^{-j\omega t} dt \\
 &= \int_{-\infty}^{\infty} \delta(t) e^{-j\omega t} dt \\
 &= \int_{-\infty}^{\infty} \delta(t+0) e^{-j\omega t} dt \\
 &= \int_{-\infty}^{\infty} \delta(t) dt = 1
 \end{aligned}$$

$$\textcircled{2} \quad e^{-at} u(t) / e^{at} u(t) \quad | \quad e^{-at} u(t) e^{-j\omega t}$$

$$\begin{aligned}
 x_1(\omega) &= \int_{-\infty}^{\infty} e^{-at} u(t) e^{-j\omega t} dt \\
 &= \int_0^{\infty} e^{-at} e^{-j\omega t} dt = \int_0^{\infty} e^{-(a+j\omega)t} dt \\
 &= \frac{1}{a+j\omega}
 \end{aligned}$$

$x_2(\omega)$  - same but  $\frac{1}{a-j\omega}$

$$\begin{aligned}
 x_3(\omega) &= x_1(\omega) + x_2(\omega) \\
 &= \frac{2a}{a^2 + \omega^2}
 \end{aligned}$$

④  $\text{sgn}(x)$

$$\lim_{t \rightarrow \infty} \int_{-\infty}^t \text{sgn}(x) dt \neq 0$$

$$\text{so, } \text{sgn}(x) = u(t) - u(-t)$$

$$X(w) = \lim_{a \rightarrow \infty} \left[ \frac{1}{a+jw} - \frac{1}{a-jw} \right]$$

$$= \lim_{a \rightarrow 0} \frac{-2jw}{a^2 + w^2} = \frac{-2j}{w} = \frac{2}{jw}$$

⑤  $u(t)$

$$u(t) = \frac{1 + \text{sgn}(x)}{2}$$

$$F(u(t)) = F\left(\frac{1}{2}\right) + \frac{1}{2}F(\text{sgn}(t))$$

$$= 2\pi\delta(w) + \frac{2}{jw} \frac{1}{2}$$

$$= \pi\delta(w) + \frac{1}{jw}$$

$$\delta(w), X = (w), X = (w), X$$

R.S.

- 10 -

(6)

$$\frac{e^{j\omega t}}{\delta(\omega - \omega_0)}$$

$$\begin{aligned}
 \textcircled{2} \quad x(t) &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \delta(\omega - \omega_0) e^{j\omega t} d\omega \\
 &= \frac{1}{2\pi} \times e^{j\omega_0 t} \int_{-\infty}^{\infty} \delta(\omega_0 - \omega) d\omega \\
 &= \frac{1}{2\pi} e^{j\omega_0 t}
 \end{aligned}$$

(7)  $\cos \omega_0 t$ 

$$\frac{1}{2} (e^{j\omega_0 t} + e^{-j\omega_0 t})$$

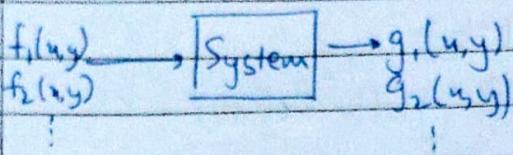
$$x(\omega) = \frac{2\pi}{\pi} [\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$$

(8)  $\text{rect}(t/T)$ 

$$\begin{aligned}
 \textcircled{2} \quad x(\omega) &= \int_{T_0}^{\infty} e^{-j\omega t} dt \Rightarrow \frac{1}{j\omega} (e^{-j\omega T_0/2} - e^{j\omega T_0/2}) \\
 &= \frac{2 \sin \frac{\omega T}{2}}{\omega} \\
 &= T \sin \frac{\omega T}{2}.
 \end{aligned}$$

## Systems

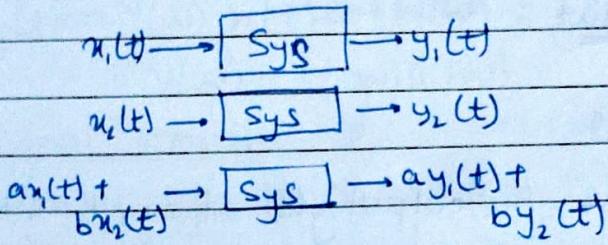
entity that processes a set of signals to yield another set of signals  
 ↳ mapping of signals



e.g. hydraulic / algorithm  
 electrical / mechanical

## Linear systems

↳ is a system which inputs linearly as in  
 if  $x_1(t)$  i/p, gives  $y_1(t)$  o/p and  $x_2$  i/p, gives  $y_2$  o/p  
 $\Rightarrow ax_1(t) + bx_2(t)$  i/p, gives  $ay_1(t) + by_2(t)$  o/p  
 where  $a, b$  are constants.



Linear systems satisfy (i) homogeneity & (ii) additivity

Eg.  $y(t) = 10x(t)$  determine linearity.

Ans.  $x_1(t) = u(t) \quad x_2(t) = \delta(t) \quad y_1(t) = 10u(t) \quad y_2 = 10\delta(t)$   
 $x(t) = ax_1(t) + bx_2(t)$   
 $y(t) = 10x(t) = 10ax_1(t) + 10bx_2(t)$   
 $= ay_1(t) + by_2(t)$

linear!

Eg.  $y(t) = x^2(t)$  determine linearity.

Ans.  $x_1 = u(t) \quad x_2 = \delta(t) \quad y_1(t) = u^2(t) \quad y_2(t) = \delta^2(t)$   
 $x(t) = ax_1(t) + bx_2(t)$   
 $y(t) = x^2(t) = a^2x_1^2(t) + b^2x_2^2(t) + 2abx_1(t)x_2(t)$   
 $\neq ay_1(t) + b(y_2(t))$

Non-linear!

## Time invariant Systems

↳ Systems whose parameters don't change with time.

$$x(t) \rightarrow \boxed{\text{Sys}} \rightarrow y(t)$$

$$x(t-t_0) \rightarrow \boxed{\text{Sys}} \rightarrow y(t-t_0)$$

Eg. ①  $y(t) = 2x(t-5)$     ②  $y(t) = 2x(3t)$

Ans. ①  $y(t) = 2x(t-5)$

$$y_1(t) = 2x_1(t-5)$$

$$x_1(t) = x_1(t-t_0)$$

$$y_2(t) - 2x_2(t-5) = 2x_2(t-5-t_0)$$

$$y_2(t-t_0) = 2x_2(t-5-t_0)$$

$$y_2(t) = y_1(t-t_0)$$

invariant

②  $y_1(t) = 2x_1(3t); y_2(t) = 2x_2(3t-t_0)$

$$y_2(t) = 2x_2(3t) = 2x_2(3t-t_0)$$

$$y_1(t-t_0) = 2x_1(3(t-t_0))$$

$$\neq y_2(t)$$

variant

## Causal Systems

↳ Systems where the output  $y(t)$  at any time  $t$  depends only on the current output  $x(t)$  at  $t$  and  $x(t-1), \dots$ . If it depends on  $x(t+1), \dots$ , it is non-causal.

i)  $y(t) = x(t) + 0.2x(t-1) \Rightarrow \text{causal}$

ii)  $y(t) = \frac{1}{3}x(t+1) + \frac{1}{3}x(t) \Rightarrow \text{non-causal}$

## Impulse Response

↳ An LTI system can be repred by its unit impulse response, which is the system's response due to the impulse input  $\delta(t)$

$$S(t) \rightarrow \boxed{\text{LTI}} \rightarrow h(t)$$

$$x(t) \rightarrow \boxed{h(t)} \rightarrow y(t)$$

$$y(t) = h(t) * x(t)$$

Q Let LTI by

$$y(t) = 0.5x(t) + 0.25x(t-1) \quad | \quad x(-1)=0$$

(a) find  $h(t)$  (b) draw block (c) write output.

Ans Let  $x(t) = \delta(t)$

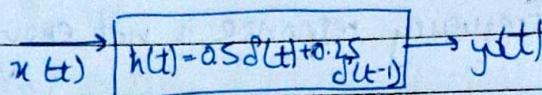
$$h(t) = 0.5x(t) + 0.25x(t-1)$$

$$h(t) = 0.5 \quad t=0$$

$$= 0.25 \quad t=1$$

$$= 0 \quad \text{else}$$

$$y(t) = h(0)x(t) + h(1)x(t-1)$$



Signal Transmission in LTI

for an LTI system

$$g(t) \xrightarrow{h(t)} y(t) \quad y(t) = h(t) * g(t)$$

$$y(t) \Leftrightarrow Y(\omega), h(t) \Leftrightarrow H(\omega) \text{ & } g(t) \Leftrightarrow G(\omega)$$

$$Y(\omega) = H(\omega)G(\omega)$$

Signal Distortion

$$|Y(\omega)|e^{j\theta_Y(\omega)} = |G(\omega)||H(\omega)|e^{j(\theta_G(\omega) + \theta_H(\omega))}$$

$$\hookrightarrow |Y(\omega)| = |G(\omega)||H(\omega)| \quad \& \quad \theta_{Y(\omega)} = \theta_{G(\omega)} + \theta_{H(\omega)}$$

Distortionless transmission

$$y(t) = G_0 x(t - t_d)$$

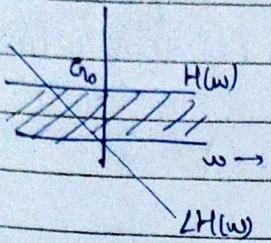
$$Y(\omega) = G_0 \cancel{\times} X(\omega) e^{-j\omega t_d}$$

$$Y(\omega) = X(\omega)H(\omega)$$

$$\hookrightarrow H(\omega) = G_0 e^{-j\omega t_d}$$

$$|H(\omega)| = G_0 \quad \angle H(\omega) = -\omega t_d$$

## Measure of time delay with freq



$$\Theta_h(w) = \omega t_d$$

$$t_d(w) = -\frac{d\Theta_h}{dw}$$

Every freq component is altered by gain & time delay of  $t_d$ .

Even flat frequency response is not enough for distortion less travel.

### Nature of audio and Video distortion

→ Humans can notice amplitude distortion, but phase dist is harder to hear

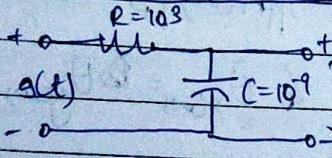
→ For phase detection to be visible,  $t_d$  has to be huge

→ Human eye is sensitive to phase distortion

→ Amplitude distortion causes very little degradation on image quality.

### Low Pass Filter Problem

Q

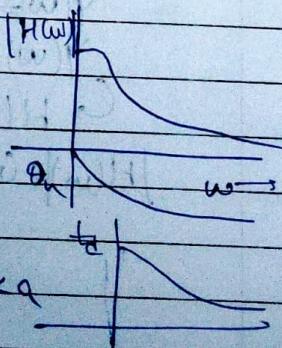


Find  $H(w)$  |  $|H(w)|$ ,  $\Theta_h(w)$

$t_d(w)$ : amp(5%) &  $\Theta(2.1)$  allowed

Ans  $H(w) = \frac{1}{jwC} = \frac{a}{R + jwC} Y(w)$   $a = (RC)^{-1}$

$$= \frac{a}{a + jw}$$



$$|H(w)| = \frac{a}{\sqrt{a^2 + w^2}} \approx 1 \text{ for } w \ll a$$

$$\Theta_h(w) = \tan^{-1} \frac{w}{a} \approx -\frac{w}{a} \text{ for } w \ll a$$

$$t_d(w) = \frac{d\Theta_h}{dw} = \frac{a}{a^2 + w^2} \approx \frac{1}{a} = 10^{-6}$$

## Nature of audio & video dist

$$|H(\omega_0)| > 0.98 \Rightarrow \frac{a}{\sqrt{a^2 + w_0^2}} > 0.98$$

$$\frac{a}{\sqrt{a^2 + w_0^2}} > 0.98 \Rightarrow w_0 < 0.203a$$

$$\frac{a}{w_0^2 + a^2} > \frac{0.95}{a} \Rightarrow w_0 < 0.2294a$$

$$t \leq \frac{1}{10^6} \quad y(t) = g t - 10^{-6}$$

## Signal energy & Energy spectral Density

### Parseval's Theorem

$$E_g = \int_{-\infty}^{\infty} g(t) g^*(t) dt = \int_{-\infty}^{\infty} g(t) \left[ \frac{1}{2\pi} \int_{-\infty}^{\infty} G_r^*(\omega) e^{-j\omega t} d\omega \right] dt$$

Q Verify for  $g(t) = e^{-at} u(t)$  ( $a > 0$ )

$$\begin{aligned} E_g &= \frac{1}{2\pi} \int_{-\infty}^{\infty} G_r^*(\omega) \left[ \int_{-\infty}^{\infty} g(t) e^{-j\omega t} dt \right] d\omega, \quad E_g = \int_{-\infty}^{\infty} g^2(t) dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} G_r(\omega) G_r^*(\omega) d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} |G_r(\omega)|^2 d\omega \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{1}{a^2 + \omega^2} d\omega = \frac{1}{2\pi} \left[ \tan^{-1} \frac{\omega}{a} \right]_{-\infty}^{\infty} = \frac{1}{2a}. \quad \text{Verified} \end{aligned}$$

### Energy Spectral Density (ESD)

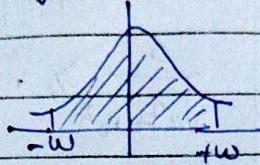
$$\begin{aligned} E_g &= \frac{1}{2\pi} \int_{-\infty}^{\infty} |G_r(\omega) H(\omega)|^2 d\omega \\ &= 2 \cdot \frac{|G_r(\omega_0)|^2}{2\pi} \int_{-\infty}^{\infty} 2 |G_r(\omega_0)|^2 d\omega \cdot \frac{1}{2\pi} \int_{-\infty}^{\infty} d\omega \quad \boxed{\frac{|G_r(\omega)|^2}{2\pi}} \end{aligned}$$

$$E_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} Y_g(\omega) d\omega = \int_{-\infty}^{\infty} \frac{|\Phi_r(\omega)|^2}{2\pi} d\omega \quad \boxed{Y_g(\omega) = |\Phi_r(\omega)|^2}$$

Q Estimate essential Bandwidth of  $g(t) = e^{-at} u(t)$  if band should contain 95% of energy.

Ans

$$G(w) = \frac{1}{1+jw}$$



$$ESD = |G(w)|^2 = \frac{1}{1+w^2}$$

$$\frac{0.95}{2a} = \frac{1}{2\pi} \int_{-w}^w \frac{dw}{w^2 + a^2} = \frac{1}{2\pi a} \tan^{-1} \frac{w}{a} \Big|_{-w}^w = \frac{1}{\pi a} \tan^{-1} \frac{w}{a}$$

$$w = 12.706a$$

Q  $g(t) = \text{rect}(\frac{t}{T})$ ; 90% band find essential bandwidth

Ans  $\text{rect}(\frac{t}{T}) \Leftrightarrow 2 \sin^2(\frac{\omega T}{2})$

$$Y_g(w) = |G(w)|^2 = T^2 \sin^2\left(\frac{\omega T}{2}\right)$$

$$E_w = \frac{1}{2\pi} \int_{-\infty}^{\infty} T^2 \sin^2\left(\frac{\omega T}{2}\right) d\omega$$

$$= \frac{1}{\pi} \int_0^{wT} \sin^2\left(\frac{x}{2}\right) dx$$

### Energy of Modulated Signal

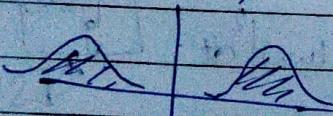
→ modulation = shifting by  $w_0$  Hz

e.g. Let  $g(t)$  = signal &  $\phi(t) = g(t)\cos w_0 t$

$$\Rightarrow \phi(w) = \frac{1}{2} [G(w+w_0) + G(w-w_0)]$$

$$\Rightarrow ESD = \frac{1}{4} [G(w+w_0) + G(w-w_0)]^2 = \frac{1}{4} [\phi]$$

if  $w_0 \gg 2\pi B$ , non-overlapping



Energy is proportional to square of amplitude, for  $g(t)$  the value is high most of the time, while  $\phi(t)$  drops to zero.

## Time autocorrelation & ESD

Cross correlation function  $\rightarrow \varphi_{g_1 g_2}(T)$

Auto correlation function  $\rightarrow \varphi_{gg}(T)$

$$ACF \Rightarrow \int_{-\infty}^{\infty} g(t) g(t+\tau) dt$$

$$\text{let } t+\tau=x \Rightarrow \int_{-\infty}^{\infty} g(u) g(x-u) du$$

$$\text{Hence ACF} \Rightarrow \int_{-\infty}^{\infty} g(t) g(t+\tau) dt$$

$\Rightarrow$  ACF is an even function as  $\varphi_g(T) = \varphi_g(-T)$

$$\text{At } \tau=0, \text{ ACF} = \int_{-\infty}^{\infty} g(t) g(t) dt = E_g$$

$\Rightarrow$  Let's find its Fourier transform [ ~~Defn of ACF~~ ]

$$\begin{aligned} F[\varphi_g(T)] &= \int_{-\infty}^{\infty} e^{-j\omega t} \left[ \int_{-\infty}^{\infty} g(u) g(u+T) du \right] d\tau \\ &= \int_{-\infty}^{\infty} g(t) \left[ \int_{-\infty}^{\infty} g(t+\tau) e^{-j\omega \tau} d\tau \right] dt \\ &= G(\omega) \int_{-\infty}^{\infty} g(t) e^{j\omega t} dt = G(\omega) G(-\omega) \\ &\quad = |G(\omega)|^2 = \varphi_g(\omega) \end{aligned}$$

$$\varphi_g(t) \Leftrightarrow \varphi_g(\omega)$$

$\Rightarrow$  ACF  $\varphi_g(t)$  is the conv of  $g(\tau)$  &  $g(-\tau)$ .

$$\begin{aligned} g(\tau) * g(-\tau) &= \int_{-\infty}^{\infty} g(u) g(-u-\tau) du \\ &= \int_{-\infty}^{\infty} g(u) g(\tau-u) du = \varphi_g(\tau) \end{aligned}$$

Ques Find the ACF of  $g(t) = e^{-at} u(t)$  also find ESD

$$\text{Ans} \quad g(t) = e^{-at} u(t) \quad g(t-\tau) = e^{-a(t-\tau)} u(t-\tau)$$

$$\varphi_g(t) = \int_{-\infty}^{\infty} g(t) g(t-\tau) d\tau = e^{-at} \int_{-\infty}^{\infty} e^{-a(t-\tau)} d\tau = \frac{1}{a} e^{-at}$$

$$\varphi_g(\omega) = \frac{1}{a^2 + \omega^2}$$

ESD of input/output

$$Y(\omega) = H(\omega)X(\omega)$$

$$|Y(\omega)|^2 = |H(\omega)|^2 |X(\omega)|^2$$

$$\Phi_y(\omega) = |H(\omega)|^2 \Phi_x(\omega)$$

O/p is  $|H(\omega)|^2$  times i/p.

Signal power

For power signal  $g(t)$ , measure is power.

$$P_g = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} g^2(t) dt \quad \left[ \text{truncate signal from } -T/2 \text{ to } T/2 \right]$$

Power Spectral Density (PSD)

$$E_g = \int_{-\infty}^{\infty} g^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |G_g(\omega)|^2 d\omega$$

$$P_g = \lim_{T \rightarrow \infty} \frac{1}{T} E_g = \lim_{T \rightarrow \infty} \left[ \frac{1}{2\pi} \int_{-\infty}^{\infty} |G_g(\omega)|^2 d\omega \right]$$

$$S_g(\omega) = \lim_{T \rightarrow \infty} \frac{|G_g(\omega)|^2}{T}$$

$$P_g = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_g(\omega) d\omega = \frac{1}{\pi} \int_0^{\pi} S_g(\omega) d\omega$$

ACF of Power Signal

$$R_g(\tau) = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} g(t)g(t+\tau) dt$$

$$= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} g(t)g(t+\tau) dt \xrightarrow{R_g(\tau) = R_g(-\tau)} \text{Hence even func}$$

Similar to last,

$$R_g(\tau) \xleftarrow{T} \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} |G_g(\omega)|^2 d\omega = S_g(\omega)$$

## Random signal

↳ A signal whose physical description is known completely by a mathematical/graphical method is called deterministic signal.

↳ A signal only known through its probabilistic description such as mean etc rather than its complete mathematical/graphical description is called random signal.

## Study Probability Theory

↳ Sample Space

↳ Basic Probability

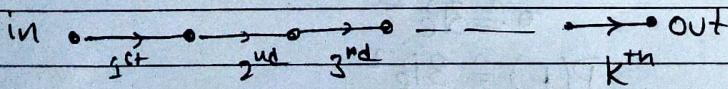
↳ Conditional Probability

↳ Bernoulli/Gaussian pdf

## PCM Repeated Error Probability

↳ regenerative repeaters are used to detect pulses and retransmit new clean pulses, this decreases pulse distortion & noise.

↳ it has  $n$  links in tandem. The pulses are detected at the end of each link and new pulse is sent over the next link. If  $P_E$  is the prob of error in detecting pulse at any link, total probability should be  $\leq n P_E$  i.e.  $P_E \leq n P_E$



Proof) One link  $\Rightarrow 1 - P_E$  [correct detection],

$n$  links  $\Rightarrow 1 - P_E^n$  [ " " ]

A pulse will be correctly detected iff ~~no~~ no error or even numbers of errors are present.

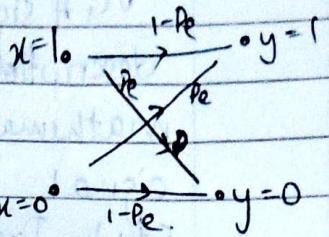
$$1 - P_E = P(\text{no error}) + P(2 \text{ error}) + P(4 \text{ error}) + \dots$$

$$P(\text{correct on all links}) = (1-P_e)^n$$

$$\& P(\text{error on } k \text{ links}) = nC_k P_e^k (1-P_e)^{n-k}$$

$$1 - P_E = (1-P_e)^n + \sum_{n=1}^{\infty} nC_n P_e^n (1-P_e)^{n-k}$$

as  $P_e \ll 1$ .



$$1 - P_E \approx (1-P_e)^n + nC_2 P_e^2$$

$$= (1-P_e)^n + \frac{n(n-1)}{2} P_e^2 \quad n=0$$

if  $nP_e \ll 1$

$$1 - P_E \approx (1-P_e)^n$$

$$= 1 - nP_e \quad [nP_e \ll 1]$$

$$\Rightarrow P_E \approx nP_e \quad \underline{\text{proved}}$$

### Binary transmission

↪ init, 0 → 00/000 and 1 → 11/111 these are ~~the~~ <sup>the</sup> norms of transmission.

if  $P_e$  is the error prob of 1 digit,  $P(E)$  is the prob of getting a wrong output

$$\text{then } P(E) = \sum_{k=2}^3 nC_k P_e^k (1-P_e)^{n-k}$$

$$= 3C_2 P_e^2 (1-P_e) + P_e^3$$

$$\approx 3P_e^2$$

$$P(E) \approx 3P_e^2$$

### Random Variables

↪ Outcomes might be real numbers

↪ some numerical values might be assigned to each sample point

↪ these variables are random variables.

## Study Distribution Function

Joint probability.

↳ prob when  $x = x_i$  &  $y = y_j$ .

$$\hookrightarrow P_{xy}(x_i, y_j)$$

$$\sum_i \sum_j P_{xy}(x_i, y_j) = 1$$

$$\sum_i P_{xy}(x_i, y_j) = \sum_i P_{xy}(x_i | y_j) P_y(y_j)$$

$$= P_y(y_j) \geq P_{xy}(x_i | y_j)$$

$= P_y(y_j)$  Marginal Probability of  $y$

$$P_x(x_i) = \sum_j P_{xy}(x_i, y_j)$$
 Marginal Probability of  $x$

Binary Symmetric Channel

↳ error probability is  $P_e$ .

↳ Prob of transmitting 1 is  $Q$  and for 0 is  $1-Q$

Determine prob of getting 0 and 1 at receiver.

Ans

$$x=0 \rightarrow y=1$$

$$P_{y|x}(0|1) = P_{y|x}(1|0) = P_e$$

$$x=0 \rightarrow y=0$$

$$P_{y|x}(1|1) = P_{y|x}(0|0) = 1-P_e$$

$$P_x(1) = Q \quad P_x(0) = 1-Q$$

$$P_y(y_j) = P_x(0) P_{y|x}(1|0) + P_x(1) P_{y|x}(0|1)$$

$$= (1-Q)P_e + Q(1-P_e)$$

Similarly,

$$P_x(x_i) = Q(1-P_e) + (1-Q)P_e$$

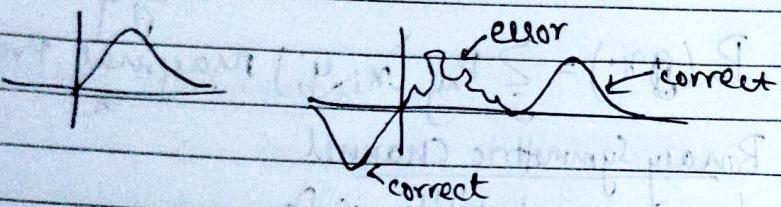
$$= QP_e + (1-Q)(1-P_e)$$

Q Prob of 0 transmitted is 0.4  
 Given  $P(E|0) = 10^{-6}$   $P(E|1) = 10^{-4}$   
 Ans  $P(E) = P_0(0)P(E|0) + P_0(1)P(E|1)$   
 $= 0.4 \times 10^{-6} + 0.6 \times 10^{-4}$   
 $= 2.604 \times 10^{-4}$

Study pdf and cdfs and gaussian pdf

Probability of Detection

↳ The received signal has the desired signal and has some random noise. This leads to pulse error detection.



In absence of noise, value of positive pulse is  $+A_p$  and else  $-A_p$ .

for additive noise, it is  $\pm A_p + n$

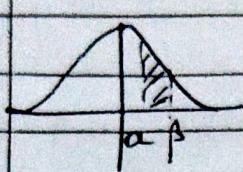
due to symmetry, the threshold is always 0.

for +ve pulse, if  $n > A_p$  and -ve, it is detected as a 0 and is wrong.

Error prob in Polar Signals

Let  $P(E|0) \Rightarrow n > A_p$

$P(E|1) \Rightarrow n < -A_p$



$$P(\alpha < n < \beta) = \frac{1}{\sigma\sqrt{2\pi}} \int_{\alpha}^{\beta} e^{-\frac{u^2}{2\sigma^2}} du.$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\alpha/\sigma}^{\beta/\sigma} e^{-u^2/2} du.$$

$$P(\varepsilon | 0) = \frac{1}{\sqrt{2\pi}} \int_{AP/\sigma_n^2}^{\infty} e^{-u^2/2} du \quad \xrightarrow{\text{Q}(u)}$$

$$\begin{aligned} P(\varepsilon) &= P(1)P(\varepsilon|1) + P(0)P(\varepsilon|0) \\ &= Q\left(\frac{AP}{\sigma_n}\right) \end{aligned}$$

Study mean, mean of sum and product

Rayleigh Density

↪ The rayleigh density is characterised by the pdf:

$$R(r) = \begin{cases} \frac{r}{\sigma^2} e^{-\frac{r^2}{2\sigma^2}} & ; r > 0 \\ 0 & ; r \leq 0 \end{cases}$$

$$P_{x|y}(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{x^2}{2\sigma^2}} \quad P_y(y) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{y^2}{2\sigma^2}}$$

$$P_{xy}(x, y) = P_x(x) P_y(y) = \frac{1}{2\pi\sigma^2} e^{-\frac{x^2+y^2}{2\sigma^2}}$$

Central Limit Theorem

↪ Under some conditions the sum of RVS take a gaussian p.d.f. The rigorous tendencies of such order is called central limit theorem.

→ Consider two RVS  $z = x + y \Rightarrow y = z - x$ .

$$\begin{aligned} F_z(z) &= P(z \leq z) = P(x \leq \infty, y \leq z-x) \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{z-x} P_{xy}(x, y) dx dy \end{aligned}$$

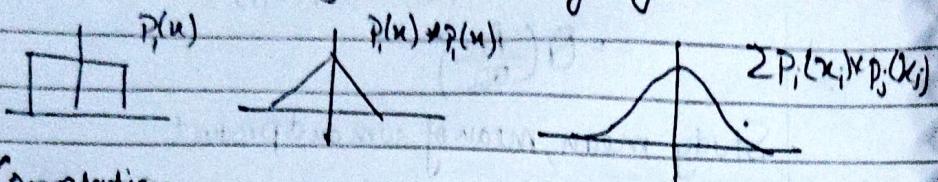
$$= \int_{-\infty}^{\infty} dx \int_{-\infty}^{z-x} P_{xy}(x, y) dy$$

$$P_z(z) = \frac{dF_z(z)}{dz} = \int_{-\infty}^{\infty} P_{xy}(x, z-x) dx$$

If  $x, y$  are independent,

$$\begin{aligned} P_{xy}(x, z-x) &= P_x(x) P_y(z-x) \\ &= \int_{-\infty}^{\infty} P_x(u) P_y(z-u) du \\ &= P_x(x) * P_y(y) \end{aligned}$$

Keep on convoluting, till inf to get gaussian curve.



### Correlation

↳ nothing but the covariance is the dependence b/w RVs.

$$\sigma_{xy} = (\bar{x} - \bar{x})(\bar{y} - \bar{y})$$

Uncorrelated  $\rightarrow \sigma_{xy} = 0$  or  $\bar{y}\bar{x} = \bar{x}\bar{y}$

Independent RVs are uncorrelated but the converse is not true.

$$\text{e.g. } g_1(t) = \sin at \quad g_2(t) = \cos at$$

### Coefficient of correlation

$$\hookrightarrow \rho_{xy} = \frac{\sigma_{xy}}{\sigma_x \sigma_y}$$

### MSE

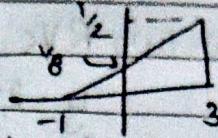
If  $x, y$  are uncorrelated,  $\bar{z} = \bar{x}\bar{y}$

$$\begin{aligned} \sigma_z^2 &= (\bar{x} - \bar{x} + \bar{y} - \bar{y})^2 \\ &= (\cancel{\bar{x}\bar{x}})(\bar{x} - \bar{x})^2 + (\cancel{\bar{y}\bar{y}})(\bar{y} - \bar{y})^2 + 2(\bar{x} - \bar{x})(\bar{y} - \bar{y}) \\ &= \sigma_x^2 + \sigma_y^2 + 2\sigma_x \sigma_y \\ &= \sigma_x^2 + \sigma_y^2. \end{aligned}$$

## Important Questions

Q Find mean, MSE, var of pdf

Aus Area should be 1



$$\bar{x} = \int_{-1}^3 x P(x) dx = \int_{-1}^3 \frac{1}{8} (y)(y-1) dy = \frac{1}{8} \left( \frac{64}{3} - \frac{16}{2} \right) = \frac{5}{3}$$

$$\bar{x}^2 = \frac{1}{8} \int_{-1}^3 x^2 (x^2+1) dx = \frac{11}{3}$$

$$\sigma_x^2 = \bar{x}^2 - \bar{x}^2 = 8/9.$$

Q Find mean, MSE and Var of n ← dice roll sum

$$\bar{x} = \sum_{x=2}^{12} x P(x) = \frac{1}{36}(2) + \frac{2}{36}(3) + \dots + \frac{1}{36}(12)$$

$$= 7.$$

$$\bar{x}^2 = \sum_{x=2}^{12} x^2 P(x) = \frac{1}{36}(4) + \frac{2}{36}(9) + \dots + \frac{1}{36}(144)$$

$$= 54.83$$

$$\sigma_x^2 = \bar{x}^2 - (\bar{x})^2 = 54.83 - 49 = 5.83.$$

Q Find mean, MSE, Var of n ← dice roll.

$$\bar{x} = \sum_{x=1}^6 x P(x) = \frac{1}{6}(1) + \frac{1}{6}(2) + \dots + \frac{1}{6}(3)$$

$$= \frac{7}{2}$$

$$\bar{x}^2 = \sum_{x=1}^6 x^2 P(x) = 391/6$$

$$\sigma_x^2 = \bar{x}^2 - (\bar{x})^2 = 305/12$$

Q Show  $\bar{z} = \bar{x} + \bar{y}$  for gaussian  $\sigma_z^2 = \sigma_x^2 + \sigma_y^2$

$$\begin{aligned} P_z(z) &= P_x(x) * P_y(y) \\ &= e^{-\frac{1}{2}(\sigma_x^2 + \sigma_y^2)w^2} e^{-j\omega(\bar{x}+\bar{y})} \\ &= \frac{1}{\sqrt{2\pi(\sigma_x^2 + \sigma_y^2)}} \end{aligned}$$

Q Proved

Q Prove  $P_{xy} \leq 1$

$$\begin{aligned} \text{Aus} \quad [a((x-\bar{x}) - (y-\bar{y}))]^2 &\geq 0 \quad \rightarrow 4\sigma_{xy}^2 - 4\sigma_x^2\sigma_y^2 \leq 1 \\ a^2\sigma_x^2 + \sigma_y^2 - 2a\sigma_{xy} &\geq 0 \quad \left| \frac{\sigma_{xy}}{\sigma_x\sigma_y} \right| = P_{xy} \leq 1 \end{aligned}$$

Q) Show two RVs  $x, y$  are

$$y = k_1 x + k_2 \epsilon$$

$P_{xy} = 1$  iff  $k_1 > 0$  and  $P_{xy} = -1$  iff  $k_1 < 0$

Ans  $\sigma_y^2 = k_1^2 \sigma_x^2$   $D_{xy} = k_1 \sigma_x^2$

$$P_{xy} = \frac{\sigma_{xy}}{\sigma_x \sigma_y} = \frac{k_1 \sigma_x^2}{k_1 \sigma_x^2} = 1 \text{ if } k_1 > 0$$

↳  $\sigma_x$  &  $\sigma_y$  are both positive,  
hence  $P_{xy} = -1$

Q

$$x = \cos \theta$$

$$y = \sin \theta$$

Show uncorrelated but not independent

Ans

$$\bar{x} = \int_0^{2\pi} \sin \theta d\theta = 0 \quad \bar{y} = \int_0^{2\pi} \cos \theta d\theta = 0.$$

$$\sigma_{xy} = (\bar{x} - \bar{x})(\bar{y} - \bar{y}) = \bar{x} \cdot \bar{y} = 0 \cdot 0 = 0$$

but  $x^2 + y^2 = 1$

$y = \sqrt{1-x^2}$  ↪ not independent.

Q

A random signal has two vals 3 & 0 with random prob  
a gaussian noise  $\chi(t)$  is added giving  $y(t)$  find pefy of  
 $y$ .

Ans

$$P_n(x) = \frac{1}{2} \delta(x) + \frac{1}{2} \delta(x-3)$$

$$P_n(u) = \frac{1}{2\sqrt{2\pi}} e^{-\frac{u^2}{8}} \quad y = x+u$$

$$P_y(y) = P_n(x) * P_n(u) = \left[ \frac{1}{2} \delta(x) + \frac{1}{2} \delta(x-3) \right] * \frac{1}{2\sqrt{2\pi}} e^{-\frac{u^2}{8}}$$

$$= \frac{1}{2} \int_{-\infty}^{\infty} \delta(x) \left[ \frac{1}{2\sqrt{2\pi}} e^{-\frac{(y-u)^2}{8}} \right] du \stackrel{\text{same}}{=} \frac{1}{2\sqrt{2\pi}} e^{-\frac{y^2}{8}}$$

$$= \frac{1}{4\sqrt{2\pi}} e^{-\frac{y^2}{8}} + \frac{1}{4\sqrt{2\pi}} e^{-\frac{(y-3)^2}{8}}$$

