



# Communication System



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# COMMUNICATION SYSTEM

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# 1

# AMPLITUDE MODULATION

## 1.1. Introduction

**Band limiting :** Bandun limited to bandlimited (LPF)

**Base band signal :** Message signals, low cut off  $f_{re} = 0$  Hz or very close to 0 Hz.

**Bandpass signal :** By shifting baseband signal to very high freq.

- Wideband signal :  $\frac{f_H}{f_L} \gg \gg 1$  (Base band signal)
- Narrowband signal :  $\frac{f_H}{f_L} \approx 1$  (Bandpass signal)

### Modulated Signal:

$$C(t) = A_c \cos(\omega_c t + \phi) = A_c \cos \omega_c t$$

Carrier signal

(carrier before modulation)

$$S(t) = A(t) \cos[\omega_c t + \phi(t)] \quad \text{Modulated signal}$$

Instantaneous  
amplitude

Instantaneous  
frequency

Instantaneous phase

### Amplitude Modulation:

DSB-FC (Double side band full carrier)

$$C(t) = A_c \cos \omega_c t \quad \text{carrier before modulation}$$

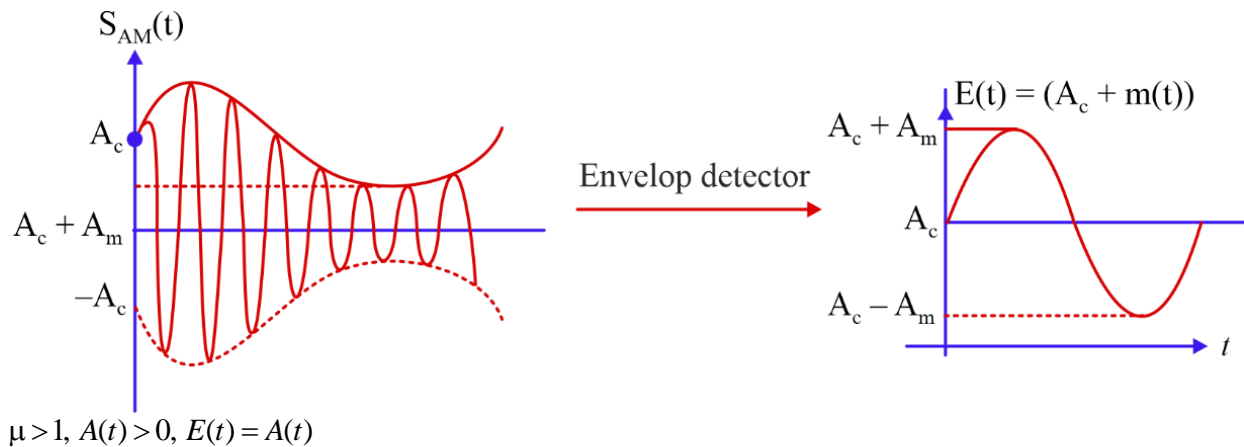
$$S_{AM}(t) = A_c \cos \omega_c t + m(t) \cos \omega_c t \Rightarrow S_{AM}(t) = [A_c + m(t)] \cos \omega_c t \quad \text{carrier after modulation}$$

$$\text{Modulation Index } \mu = \frac{[m(t)]}{A_c} \max$$

(1)  $\mu < 1$  (under modulation)

$$\mu = \frac{A_m}{A_c} < 1$$

$$\mu = \frac{[E(t)]_{\max} - [E(t)]_{\min}}{[E(t)]_{\max} + [E(t)]_{\min}}$$



$$\mu > 1, A(t) > 0, E(t) = A(t)$$

Recovery through E, D possible.

$$S(t)_{\max} = E(t)_{\max} = A_c(1 + \mu)$$

$$S(t)_{\min} = E(t)_{\min} = A_c(1 - \mu)$$

- (2) Critical Modulation:-  $\mu = 1, A(t) \geq 0, E(t) = A(t)$ ,  $m(t)$  can be recovered with envelope detector.
- (3) Over modulation:  $\mu > 1, A(t) \not\geq 0, E(t) = |A(t)|$ , not possible by E.D

### Frequency Related Parameters

- (1)  $m(t) \rightarrow \begin{cases} \text{B.W} = f_m \\ f_{\max} = f_m \end{cases}$
- (2)  $C(t) \rightarrow f_{\max} = f_m$
- (3)  $S(t) \rightarrow \begin{cases} \text{B.W} = 2f_m, f_{\max} = f_c + f_m \\ f_{\min} = f_c - f_m \end{cases}$  (2 times of freq. of  $m(t)$ )

$$\Rightarrow P_{AM} = P_C + P_{SB}$$

$$P_{USB} = P_{LSB} = \frac{P_m}{4}$$

### Modulation efficiency

$$\eta = \frac{P_{SB}}{P_{AM}} = \frac{P_m / 2}{P_c + \frac{P_m}{2}}$$

Share of sideband power in total power

- $$\Rightarrow k_a \text{ [Amplitude sensitivity of amplitude modulator]}$$

$$k_a = \frac{1}{A_c} \text{ (per volt),}$$

$$A(t) = A_c[1 + k_a m(t)]$$

$A(t) > 0$ , E. D. Applicable

$$\text{DSB - FC} \rightarrow \begin{cases} [A_c + m(t)] \cos 2\pi f_c t \rightarrow \mu = \frac{|m(t)|_{\max}}{A_c} \\ A_c [1 + k_a m(t)] \cos 2\pi f_c t \rightarrow \mu = k_a |m(t)|_{\max} \end{cases}$$

### For single tone sinusoidal signal

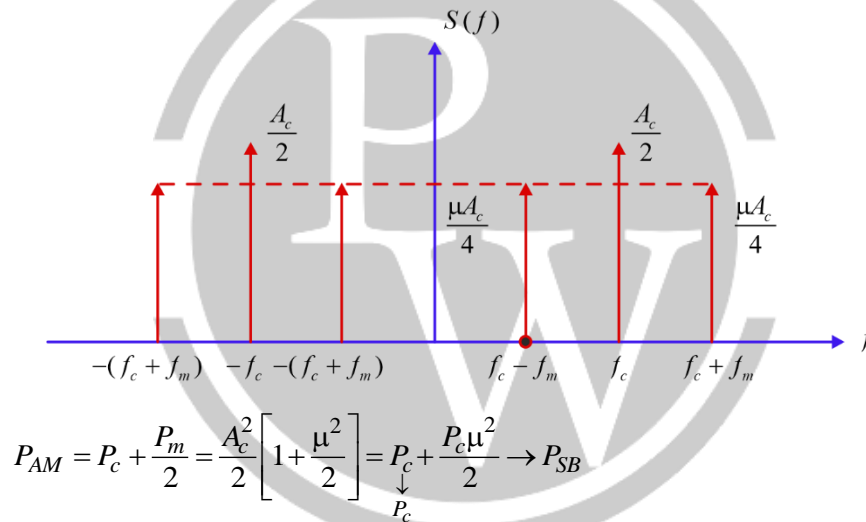
$$f_{\max} = f_m, BW = 0\text{Hz}, P_m = \frac{A_m^2}{2} \rightarrow \text{for message signal}$$

$$\mu = \frac{A_m}{A_c}$$

$$\begin{aligned} S_{AM}(t) &= [A_c + A_m \cos 2\pi f_m t] \cos 2\pi f_c t \\ &= A_c [1 + \mu \cos 2\pi f_m t] \cos 2\pi f_c t \end{aligned}$$

$$S_{AM}(t) = A_c \cos 2\pi f_c t + \frac{\mu A_c}{2} \cos[2\pi(f_c + f_m)t] + \frac{\mu A_c}{2} \cos[2\pi(f_c - f_m)t]$$

$\downarrow$  carrier
 $\downarrow$  USB
 $\downarrow$  LSB



$$\eta = \frac{\frac{P_c \mu^2}{2}}{P_c + \frac{P_c \mu^2}{2}} \Rightarrow \% \eta = \frac{\mu^2}{2 + \mu^2} \times 100\%$$

➤ If  $k_a$  given  $\rightarrow \mu = k_a |m(t)|_{\max}$

If  $k_a$  not given  $\rightarrow \mu = \frac{|m(t)|_{\max}}{A_c}$

### Important Points:

$\mu = 0$	$\mu = 1$	% Change
$P_{AM} = P_c$	$P_{AM} = 1.5P_c$	50 %
$\eta = 0$	$\eta = \frac{1}{3} = 33.33 \%$	0 % to 33.33 %

$$(2) \mu \uparrow \rightarrow \eta \uparrow$$

$$(3) P_{AM} \rightarrow \text{Will be constant if } P_c \uparrow \text{ and } \mu \downarrow$$

If  $m(t)$  is multiple single tone signal-

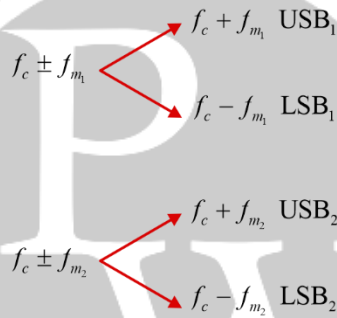
$$S(t) = A_c \left[ 1 + \frac{A_{m1}}{A_c} \cos 2\pi f_{m1} t + \frac{A_{m2}}{A_c} \cos 2\pi f_{m2} t \right] \cos 2\pi f_c t$$

$$\mu_1 = \frac{A_{m1}}{A_c}, \mu_2 = \frac{A_{m2}}{A_c} - \mu_1 > \mu_2$$

$$m(t) \rightarrow f_{m1}, f_{m2} \rightarrow f_{\max} = f_{m2}$$

$$f_{m2} > f_{m1} \quad BW = f_{m2} - f_{m1}$$

$$S(t) = f_c \quad BW = 2 \times \text{Max. Freq. component of } m(t)$$



### Power Related Parameters

$$P_m = \frac{A_{m1}^2}{2} + \frac{A_{m2}^2}{2}$$

$$P_{AM} = P_c + \frac{P_m}{2} = \frac{A_c^2}{2} \left[ 1 + \frac{\mu_1^2}{2} + \frac{\mu_2^2}{2} \right]$$

$$P_{USB1} = P_{LSB1} = \frac{P_c \mu_1^2}{2}, P_{USB2} = P_{LSB2} = \frac{P_c \mu_2^2}{2}$$

$$P_{USB1} = P_{LSB1} = \frac{P_c \mu_1^2}{4}, P_{USB2} = P_{LSB2} = \frac{P_c \mu_2^2}{4}$$

$$\eta = \frac{P_{SB}}{P_{AM}} = \frac{\mu_T^2}{2 + \mu_T^2}$$

$$\mu_T = \mu_1^2 + \mu_2^2 + \mu_3^2 + \dots$$

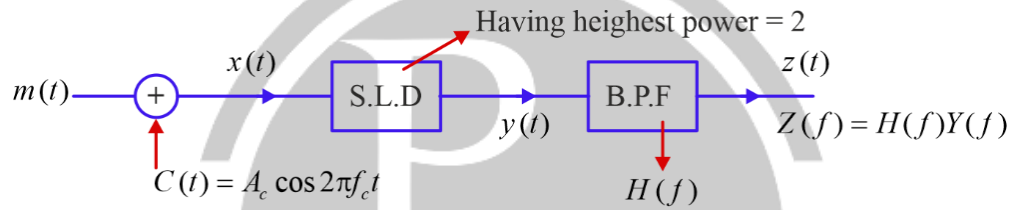
**Important Points:**

	$m(t)$ (volt)	$P_{AM}$	$P_{rod}$
(1)	Sinusoidal	$P_c \left( 1 + \frac{\mu^2}{2} \right)$	$\frac{P_c}{R} \left( 1 + \frac{\mu^2}{2} \right)$
(2)	Square wave	$P_c (1 + \mu^2)$	$\frac{P_c}{R} (1 + \mu^2)$
(3)	Triangular wave	$P_c \left( 1 + \frac{\mu^2}{3} \right)$	$\frac{P_c}{R} \left( 1 + \frac{\mu^2}{3} \right)$

$$V_{AM} = V_c \sqrt{1 + \frac{\mu^2}{2}}, \quad I_{AM} = I_c \sqrt{1 + \frac{\mu^2}{2}} \text{ for sinusoidal}$$

**DSB-FC [AM] Modulator**

**(1) Square law Modulator:**



$$y(t) = \underbrace{a_0 m(t)}_{(1)} + \underbrace{a_0 A_c \cos 2\pi f_c t}_{(2)} + \underbrace{a_1 m^2(t)}_{(3)} + \underbrace{\frac{a_1 A_c^2}{2}}_{(4)} + \underbrace{\frac{a_1 A_c^2}{2} \cos 4\pi f_c t}_{(5)} + \underbrace{2m(t)A_c \cos 2\pi f_c t}_{(6)}$$

➤ Only (2) and (6) are desirable

$$Z(t) = a_0 A_c \cos 2\pi f_c t \left[ 1 + \frac{2a_1}{a_0} m(t) \right] \text{ DSB-FC}$$

$$Z(t) = A_c' [1 + k_a m(t)] \cos 2\pi f_c t \text{ only when } f_c \gg \gg 3f_m, f_c \gg \gg (2+1)f_m$$

$$A_c' = a_0 A_c, \quad k_a = \frac{2a_1}{a_0}, \quad \mu = k_a |m(t)|_{\max}$$

**(2) Switching Modulator:**

$$Z(t) = \frac{A_c}{2} \left[ 1 + \frac{4}{\pi A_c} m(t) \cos 2\pi f_c t \right] \text{ DSB-FC}$$

$$Z(t) = A_c' [1 + k_a m(t)] \cos 2\pi f_c t$$

$$A_c' = \frac{A_c}{2}, \quad k_a = \frac{4}{\pi A_c}, \quad \mu = k_a |m(t)|_{\max}$$



## DSB- FC Demodulator-

### (1) Square law demodulator-

$$Y(t) = a_1 A_c A_m \cos 2\pi f_m t + \frac{a_1 A_m^2}{4} + \frac{a_1 A_m^2}{4} \cos 4\pi f_m t$$

$$Y(t) = B_0 + B_1 \cos \omega_0 t + B_2 \cos 2\omega_0 t$$

➤ 2<sup>nd</sup> harmonic distortion  $D_2 = \left| \frac{B_2}{B_1} \right| = \frac{\mu}{4}$

$$(D_2)_{\max} \% = 25\%$$

➤ Practically not used

➤  $\left( \frac{S}{I} \right)_{\min} = \frac{2}{\mu}$

### Envelope Detector:

$$\sqrt{A^2 + B^2} \cos \omega_c t \rightarrow \boxed{E.D} \rightarrow \sqrt{A^2 + B^2}$$

(1)  $x(t) = A \cos \omega_0 t + B \sin \omega_0 t \rightarrow E(t) = \sqrt{A^2 + B^2}$

(2)  $x(t) = A \cos(\omega_0 t + \theta) + B \sin \omega_0 t \rightarrow E(t) = \sqrt{A^2 + B^2 - 2AB \sin \theta}$

(3)  $x(t) = A(t) \cos \omega_c t \rightarrow E(t) = |A(t)|$

(4)  $x(t) = (A_c + m(t)) \cos \omega_c t \rightarrow E(t) = |A_c + m(t)|$

### Important Points:

➤ Used only when  $\mu \leq 1$

➤  $T_c = R_S C \ll \frac{1}{f_c}$  (charging time constant)

➤  $T_c = R_S C \gg \frac{1}{f_c} \rightarrow$  Peaks are not detected.

➤ Diagonal clipping  $\rightarrow R_L C = \frac{1}{f_m}$

➤ To avoid diagonal clipping  $R_L C \ll \frac{1}{f_m}, R_L C \leq \frac{\sqrt{1-\mu^2}}{\omega_m \mu}$

➤  $T_a = R_L C \approx \frac{1}{f_c}$  fluctuation is output

➤ To remove fluctuation  $R_L C \gg \frac{1}{f_c}$

➤ Proper choice of discharging time constant  $R_L C$  -

No fluctuation  $\left( \frac{1}{f_c} \lll R_L C \lll \frac{1}{f_m} \right)$  No diagonal clipping

(7)  $m(t)$ : Multitone  $f_m \rightarrow f_{\max} = \text{Max. freq. component of } m(t)$

## 1.2. Synchronous Detector

$\Delta\omega$	$\Delta\phi$	$\Delta\phi$	Recovery
$= 0$	$\neq 0$	$\pm(2n+1)\frac{\pi}{2}$	✓
$= 0$	$\neq 0$	$= (2n+1)\frac{\pi}{2}$	Q.N.E
$\neq 0$	$= 0$	$= 0$	×
$= 0$	$= 0$	$= 0$	✓

### DSB-SC :

$$S_{DSB-SC}(t) = m(t)A_c(\cos 2\pi f_c t) \quad E(t) = |A(t)| = A_c m(t)$$

➤  $B.W = 2 \times \text{max. freq. component of } m(t)$

$$P_{DSB} = P_m P_c = P_{SB} \rightarrow P_{USB} = P_{LSB} = \frac{P_{SB}}{2} = \frac{P_m P_c}{2}$$

$$P_{DSB} = \frac{P_c \mu^2}{2} = P_{SB}$$

➤ Single tone modulation.

$$S(t) = \frac{A_c A_m}{2} \cos[2\pi(f_c + f_m)t] + \frac{A_c A_m}{2} \cos[2\pi(f_c - f_m)t]$$

$$P_{DSB} = P_m P_c = \frac{A_c^2 A_m^2}{4}$$

$$\text{Multiline } P_{DSB} = P_c P_m = \frac{A_c^2}{2} \left[ \frac{A_{m1}^2}{2} + \frac{A_{m2}^2}{2} \right]$$

$$\text{Square wave - } P_{DSB} = P_c P_m = \left( \frac{A_c^2}{2} \right) A_m^2$$

$$\text{Triangular wave } P_{DSB} = P_c P_m = \left( \frac{A_c^2}{2} \right) \left( \frac{A_m^2}{3} \right) a$$

$$\text{Saw-toothed wave- } P_{DSB} = P_c P_m = \left( \frac{A_c^2}{2} \right) \left( \frac{A_m^2}{3} \right)$$

(1) Balanced Modulator-  $S_{DSB}(t) = 2A_c k_a m(t) \cos f_c t$

(2) Ring Modulator-  $y(t) \propto m(t) \cos \omega_c t$

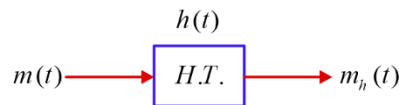
$$\Delta\omega = 0, \Delta\phi \neq 0, y(t) = 0 \quad \text{QNE}$$

$$\Delta\omega \neq 0, \Delta\phi = 0, y(t) = \frac{A_c A_c'}{2} m(t) \cos(\Delta\omega t) \rightarrow \text{distorted } m(t)$$

$$\Delta\omega = 0, \Delta\phi = 0, y(t) = \frac{A_c A_c'}{2} m(t) \rightarrow \text{Attenuated}$$

### Hilbert Transformation.

$$h(t) = \frac{1}{\pi t},$$



$$mh(t) = m(t) * \frac{1}{\pi t}$$

$$H(\omega) = -j \operatorname{sgn}(\omega) = \begin{cases} -j & \omega > 0 \\ 0 & \omega = 0 \\ j & \omega < 0 \end{cases}$$

- $M_h(f) = M(f)[-j \operatorname{sgn}(f)]$
- $H.T[\cos \omega(t)] = \sin \omega(t) \xrightarrow{H.T.} -\cos \omega(t)$
- Non causal LTI system.
- $x(t) \xleftarrow{H.T.} x_h(t)$   
 $x_h(t) \xleftarrow{H.T.} -x(t)$
- Magnitude spectrum of  $x(t)$  and  $x_h(t)$  will be same
- If  $x(t)$ : Band limited then  $x_h(t)$  is also bandlimited.
- If  $x(t)$  is non periodic then  $x_h(t)$  is also non periodic
- $x(t)$  and  $x_h(t)$  are orthogonal signal.

### Drawback of DSB-SC

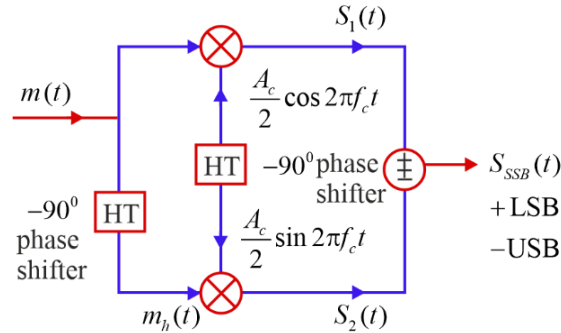
- 2 sideband Txed.
- If receiver is designed in such a way that it may recover the complete message signal from single SB then DSB-SC S/S becomes impractical.

### SSB- SC (Single sideband suppressed carrier)

- (1) Point to point communication
- (2) Two methods of generation
  - Phase dersemination
  - Frequency discrimination

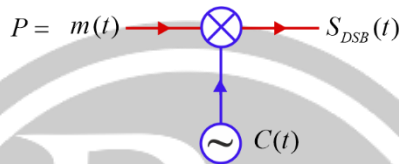
(a) Phase Discrimination:

$$S(t)_{SSB} = \frac{A_c m(t)}{2} \cos 2\pi f_c t \pm \frac{A_c m_h(t)}{2} \sin 2\pi f_c t \begin{matrix} + \Rightarrow LSB \\ - \Rightarrow USB \end{matrix}$$



Problem Solving

- (1) Identify the phase discrimination setup
- (2) Phase discrimination setup
- (3) Phase discrimination setup:



$$+ \Rightarrow S_{SSB}(t) \rightarrow LSB$$

$$- \Rightarrow S_{SSB}(t) \rightarrow USB$$

Spectral gap in D.S.B	BPF	Signal
0 Hz	Ideal	SSB-SC
0 Hz	Practical	VSB-SC
$\neq 0$ Hz	Ideal	SSB -SC
$\neq 0$ Hz	Practical	depends on practical BPF $\begin{cases} V_{SB} - SC \\ SSB - SC \end{cases}$

➤ SSB- SC can be demodulated by Synchronous detection.

- (1)  $\Delta\omega = 0, \Delta\phi \neq 0, m(t)$  recovery not possible  $\rightarrow$  freq. synchronization
- (2)  $\Delta\omega \neq 0, \Delta\phi = 0, m(t)$  recovery not possible  $\rightarrow$  Phase synchronization
- (3) Perfect syne,  $\Delta\phi = 0, \Delta\omega = 0$  can be recovered
- (4)  $\Delta\omega = 0, \Delta\phi = \frac{\pi}{2} \rightarrow$  No QNE

Note:

- (1) When video signal is transmitted through SSB- SC modular VSB- SC is generated.
- (2) Synchronous detector can not recover  $m(t)$  video signal from the above generated VSB- SC.

### Percentage Power Saved

(1) % power saved in DSB- SC as compare to DSB-FC.

$$\% P_{\text{saved}} = \frac{P_{\text{saved}}}{P_{\text{Total}}} \times 100\%$$

$$\% P_{\text{saved}} = \frac{P_c}{P_c \left[ 1 + \frac{\mu^2}{2} \right]} = \frac{2}{2 + \mu^2} = (1 - \eta)$$

(2) % power saved in SSB- SC as compare to DSB-FC-

$$\% P_{\text{saved}} = \frac{4 + \mu^2}{4 + 2\mu^2}$$

(3) % power saved in SSB-SC as compared to DSB-SC.

$$\% P_{\text{saved}} = 50\%$$

Modulation		B.W	Power	Application
(1)	DSB-FC	$2f_{\text{max}}$	$P_C + P_{SB}$	Broadcasting
(2)	DSB- SC	$2f_{\text{max}}$	$P_{SB}$	×
(3)	SSB-SC	$f_{\text{max}}$	$\frac{P_{SB}}{2}$	Point to point voice communication
(4)	VSB-SC	$f_{\text{max}} < f < 2f_{\text{max}}$	$\frac{P_{SB}}{2} < P_{VSB} < P_{SB}$	Point to point video communication.

### Pre envelope and Complex Envelope

(1) Pre Envelope calculated for both baseband and bandpass signal.

Let  $x(t)$  is real signal.

$$x_+(t) = \text{Pre envelope of } x(t)$$

$$x_+(t) = x(t) + j\hat{x}(t)$$

$$\hat{x}(t) = HT [x(t)]$$

$$x_+(f) = x(f)[1 + \text{sgn}(f)]$$

**Complex Envelope:** For bandpass only but result in low pass only

$x(t) \rightarrow$  Bandpass signal.

**Step-1.** Calculate  $x_+(t) = x(t) + j\hat{x}(t)$

**Step 2.**  $\frac{x_c(t) = x_+(t)e^{-j\omega_c t}}{X_c(f) = X_+(f + f_c)}$  left shift of pre envelope by  $f_c$



# 2

## ANGLE MODULATION

### 2.1. Introduction

	Signal = $x(t)$	$ x(t) _{\max}$
(1)	$A \cos \omega_0 t + B \cos \omega_0 t$	$ A + B $
(2)	$A \sin \omega_0 t + B \sin \omega_0 t$	$ A + B $
(3)	$A \sin \omega_0 t + B \cos \omega_0 t$	$\sqrt{A^2 + B^2}$
(4)	$A \cos \omega_1 t + B \cos \omega_2 t$	$ A + B $
(5)	$A \sin \omega_1 t + B \sin \omega_2 t$	$ A + B $
(6)	$A \cos \omega_1 t + B \sin \omega_2 t$	$\begin{cases} \rightarrow  A + B  \text{ if } A = B \\ \rightarrow <  A + B  \text{ if } A \neq B \end{cases}$

#### 2.1.1. Instantaneous Angle and Instantaneous frequency-

$$S(t) = A_c \cos[\theta_i(t)]$$

$\theta_i(t) \rightarrow$  Instantaneous angle (rad)

$$\frac{d\theta_i(t)}{dt} = \omega_i(t) \rightarrow \text{instantaneous angular frequency.}$$

$$f_i(t) = \frac{1}{2\pi} \frac{d\theta_i(t)}{dt} \text{ or } f_i(t) = \frac{\omega_i(t)}{2\pi}$$

$$\theta_i(t) = \int_{-\infty}^t \omega_i(t) dt$$

- Angle Modulation :**

- Frequency Modulation
- Phase Modulation

- Frequency Modulation :**

$$S_{angle}(t) = A_c \cos[\omega_c t + \Delta\phi(t)]$$

If angle Modulation is FM,  $\frac{d\Delta\phi(t)}{dt} \propto m(t)$

$$\frac{d\Delta\phi(t)}{dt} = K_f m(t), \quad K_f = \text{frequency sensitivity of frequency modulator}$$

$$\omega_i(t) = \omega_c + K_f m(t) \Rightarrow \omega_i(t) = \omega_c + \underset{\substack{\downarrow \\ \text{frequency} \\ \text{deviation}}}{\Delta\omega(t)}$$

$$\theta_i(t) = \theta_c + \int_{-\infty}^t K_f m(t) dt \quad \Delta\omega(t) = K_f m(t)$$

$$\Delta\omega(t) = \frac{d\Delta\phi(t)}{dt}$$

### Few Important Results

For Important Results	For $K_f : \frac{\text{rad}}{\text{V} \cdot \text{sec}}$	$K_f : \frac{\text{Hz}}{\text{Volt}}$
1. Instantaneous frequency	$\omega_i(t) = \omega_c + K_f m(t)$	$f_i(t) = f_c + K_f m(t)$
2. Instantaneous frequency deviation	$\Delta\omega(t) = K_f m(t)$	$\Delta f(t) = K_f m(t) \text{ Hz}$
3. Frequency deviation in +ve direction	$[\Delta\omega(t)]_{\max} = K_f [m(t)]_{\max}$	$[\Delta f]_{\max} = K_f [m(t)]_{\max}$
4. Frequency deviation in -ve direction	$[\Delta\omega(t)]_{\min} = K_f [m(t)]_{\min}$	$[\Delta f(t)]_{\min} = K_f [m(t)]_{\min}$
5. Maximum value of instantaneous frequency	$[\omega_i(t)]_{\max} = \omega_c + [\Delta\omega(t)]_{\max}$	$[f_i(t)]_{\max} = f_c + [\Delta f(t)]_{\max}$
6. Minimum value of instantaneous frequency	$[\omega_i(t)]_{\min} = \omega_c + [\Delta\omega(t)]_{\min}$	$[f_i(t)]_{\min} = f_c + [\Delta f(t)]_{\min}$
7. Peak to peak frequency deviation	$[\Delta\omega]_{p-p} = [\omega_i(t)]_{\max} - [\omega_i(t)]_{\min}$	$[\Delta f]_{p-p} = [f(t)]_{\max} - [f(t)]_{\min}$
8. Maximum frequency deviation		
9. Modulation index or deviation ratio of FM	$ \Delta\omega(t) _{\max} = K_f [m(t)]_{\max}$	$ \Delta f(t) _{\max} = K_f [m(t)]_{\max}$
$B_{FM} = \frac{\text{Maximum frequency deviation}}{\text{Maximum frequency component of } m(t)}$	$B_{FM} = \frac{K_f  m(t) _{\max}}{\omega_{\max}}$	$B_{FM} = \frac{K_f  m(t) _{\max}}{f_{\max}}$

Important Phase Calculation	$K_f \left( \frac{\text{rad}}{\text{V} \cdot \text{sec}} \right)$	$K_f : \frac{\text{Hz}}{\text{Volt}}$
1. Instantaneous phase deviation in FM	$\Delta\phi(t) = K_f \int_{-\infty}^t m(\tau) d\tau$	$\Delta\phi(t) = 2\pi K_f \int_{-\infty}^t m(\tau) d\tau$
2. Maximum phase deviation in FM	$ \Delta\phi(t) _{\max} = K_f \left  \int_{-\infty}^t m(\tau) d\tau \right $	$2\pi K_f \left  \int_{-\infty}^t m(\tau) d\tau \right _{\max}$

### General expression for FM

$$K_f : \frac{\text{rad}}{\text{V-sec}}$$

$$S_{angle}(t) = A_c \cos \left[ \omega_c t + \int_{-\infty}^t K_f m(\tau) d\tau \right]$$

$$\text{For } K_f : \frac{\text{Hz}}{\text{Volt}}$$

$$S_{FM} = A_c \cos \left[ \omega_c t + 2\pi K_f \int_{-\infty}^t m(\tau) d\tau \right]$$

$$\text{For } m(t) = A_m \cos 2\pi f_m t -$$

$$f_{\max} = f_m, [m(t)]_{\max} = +A_m, [m(t)]_{\min} = -A_m, |m(t)|_{\max} = A_m$$

$$S_{FM}(t) = A_c \cos [\omega_c(t) + B_{FM} \sin(2\pi f_m t)]$$

$$\text{For } m(t) = A_{m_1} \cos 2\pi f_{m_1} t + A_{m_2} \cos 2\pi f_{m_2} t -$$

$$f_{\max} = (f_{m_1}, f_{m_2})_{\max}$$

$$S_{FM}(t) = A_c \cos [\omega_c(t) + B_1 \sin 2\pi f_{m_1} t + B_2 \sin 2\pi f_{m_2} t]$$

$$B_1 = \frac{K_f A_{m_1}}{f_{m_1}}, B_2 = \frac{K_f A_{m_2}}{f_{m_2}}$$

### Phase Modulation –

$$\Delta\phi(t) \propto m(t)$$

$$\begin{matrix} \Delta\phi(t) = K_p m(t) \\ \downarrow \quad \quad \downarrow \\ \text{rad} \quad \quad \text{Volt} \end{matrix}$$

$K_p$  : Phase sensitivity of phase modulator

$$K_p = \frac{\text{rad}}{\text{Volt}}$$

### Phase Calculation :

$$K_p : \text{rad/Volt}$$

$$\theta_i(t) = \omega_c t + K_p m(t)$$

1. Instantaneous phase deviation  $= \Delta(t) = K_p m(t)$
2. Maximum phase deviation  $= |\Delta\phi(t)|_{\max} = K_p |m(t)|_{\max}$



### Frequency Calculation

$$\omega_i(t) = \omega_c + \Delta\omega(t)$$

$$\triangleright \Delta\omega(t) = K_p \frac{dm(t)}{dt}$$

$$\triangleright |\Delta\omega(t)|_{\max} = K_p \left| \frac{dm(t)}{dt} \right|_{\max}$$

$$\triangleright |\Delta\omega(t)|_{\min} = K_p \left| \frac{dm(t)}{dt} \right|_{\min}$$

$$\triangleright |\Delta\omega_i(t)|_{\max} = \omega_c + [\Delta\omega(t)]_{\max}$$

$$\triangleright |\Delta\omega_i(t)|_{\min} = \omega_c + [\Delta\omega(t)]_{\min}$$

$$\triangleright \Delta\omega_{p-p} = [\omega_i(t)]_{\max} - [\omega_i(t)]_{\min}$$

$$\triangleright |\Delta\omega(t)|_{\max} = K_p \left| \frac{dm(t)}{dt} \right|_{\max}$$

$$\triangleright \beta_{FM} = \frac{|\Delta\omega(t)|_{\max}}{\omega_{\max}} = \frac{K_p \left| \frac{dm(t)}{dt} \right|_{\max}}{\omega_{\max}}$$

$$S_{FM}(t) = A_c \cos[\omega_c(t) + K_{PM}(t)]$$

When  $m(t) = A_m \cos 2\pi f_m t$

$$|m(t)|_{\max} = A_m, f_{\max} = f_m, \Delta\omega(t) = -K_p A_m \omega_m \sin \omega_m t$$

$$\triangleright [\Delta\omega(t)]_{\max} = K_p A_m \omega_m$$

$$\triangleright \{\omega(t)\}_{\min} = -K_p A_m \omega_m$$

$$\triangleright [\omega_i(t)]_{\max} = \omega_c + K_p A_m \omega_m, [\omega_i(t)]_{\min} = \omega_c - K_p A_m \omega_m$$

$$\triangleright (\Delta\omega)_{p-p} = 2K_p A_m \omega_m$$

$$\triangleright |\Delta\omega(t)|_{\max} = K_p A_m \omega_m$$

$$\triangleright \beta = K_p A_m = |\Delta\phi(t)|_{\max}$$

$$S_{PM}(t) = A_c \cos[\omega_c(t) + \beta_{PM} \cos 2\pi f_m t]$$

$$S_{PM}(t) = A_c \cos[\omega_c(t) + \beta_1 \cos 2\pi f_{m_1} t + \beta_2 \cos 2\pi f_{m_2} t]$$

### Types of FM –

- Narrow Band ( $\beta \ll 1$ )
- Wide Band

$$S_{FM}(t) = A_c \cos 2\pi f_c t + \frac{A_c \beta}{2} \cos[2\pi(f_c + f_m)t] - \frac{A_c \beta}{2} \cos[2\pi(f_c - f_m)t]$$

↓ Carrier
↓ USB
↓ LSB

$$S_{FM}(t) = S_{NBFM}(t)$$

- B.W =  $2f_m$
- $P_{NBFM} = P_C \left(1 + \frac{\beta^2}{2}\right)$   $\beta \ll 1, \beta^2 \ll 1$

$$P_{NBFM} \approx P_C = \frac{A_c^2}{2}$$

### Relation between DSB-FC and NBFM –

$$S_{AM}(t) = A_c \cos 2\pi f_c t + \frac{A_c \mu}{2} \cos[2\pi(f_c + f_m)t] + \frac{A_c \mu}{2} \cos[2\pi(f_c - f_m)t]$$

$$S_{NBFM}(t) = A_c \cos 2\pi f_c t + \frac{A_c \beta}{2} \cos[2\pi(f_c + f_m)t] - \frac{A_c \beta}{2} \cos[2\pi(f_c - f_m)t]$$

1.

Frequency Component	Strength AM	Strength NBFM
$f_c$	$\frac{A_c}{2}$	$\frac{A_c}{2}$
$f_c + f_m$	$\frac{\mu A_c}{4}$	$\frac{\beta A_c}{4}$
$f_c - f_m$	$\frac{\mu A_c}{4}$	$\frac{-\beta A_c}{4}$

2.  $S_{NBFM}(t) + S_{AM}(t) = \text{SSB-SC} \rightarrow \text{USB-FC}$

$S_{AM}(t) - S_{NBFM}(t) = \text{SSB-SC} \rightarrow \text{LSB-FC}$

3. LSB in NBFM is  $180^\circ$  inverted w.r.t to LSB in AM

- $S_{FM}(t) = A_c \sum_{n=-\infty}^{\infty} J_n(\beta) \cos[2\pi(f_c + n f_m)t]$

For any value of  $\beta$

- $J_n(\beta) = (-1)^n J_n(\beta)$

$$\sum_{n=-\infty}^{\infty} J_n^2(\beta) = 1$$

$$J_0(\beta) = 0, \beta = 2.4, 5.5, 8.6, 11.8$$

$$\text{as } n \uparrow, \rightarrow J_n(\beta) \downarrow$$

$$\beta \ll 1: S(t) \rightarrow 1 \text{ Carrier} + 2 \text{ SB} \quad \text{NB Angle Modulation}$$

$$\text{If } \beta \gg 1: S(t): 1 \text{ Carrier} + \text{Infinite SB} \quad \text{Wide Band Angle Modulation}$$

$$\text{Ideal BW of WBFM} = \infty$$

### Carson's Rule –

$$BW = (\beta + 1) 2f_m \quad \beta_{PM} \text{ for PM}$$

$$\beta_{FM} \text{ for FM}$$

$$\text{Power of Carrier before modulation} = \frac{A_c^2}{2} = P_c$$

$$\text{Power of Carrier after modulation} P = P_c [J_0^2(\beta) + 2(J_1^2(\beta) + J_2^2(\beta) + \dots)]$$

$$\text{Power of Carrier component in modulated signal} = P_c J_0^2(\beta)$$

$$P_{SB} = 2P_c [J_1^2(\beta) + J_2^2(\beta) + \dots]$$

$$\eta = \frac{P_{SB}}{P_{Total}} \rightarrow \frac{P_{SB}}{P_c [J_0^2(\beta) + 2(J_1^2(\beta) + J_2^2(\beta) + \dots)]}$$

$$\text{If } J_0(\beta) = 0 \text{ then } \eta = 100\%$$

$$\text{For Infinite sidebands } P_{WB} = P_c$$

### For Non sinusoidal –

$$S_{FM} = A_c \sum_{n=-\infty}^{\infty} |C_n| \cos[2\pi(f_c + nf_m)t + \angle C_n]$$

$$m(t) \quad \text{BW}$$

$$\text{Single tone sinusoidal} \longrightarrow (\beta + 1) 2f_m$$

$$\text{Non sinusoidal} \longrightarrow (\beta + 1) 2f_m, f_m = \text{fundamental frequency}$$

$$\text{periodic signal}$$

$$\text{Other Cases} \longrightarrow (\beta + 1) 2f_{\max}$$

$$BW = (1 + \beta) 2f_m \text{ or } 2(\Delta f + f_{\max})$$

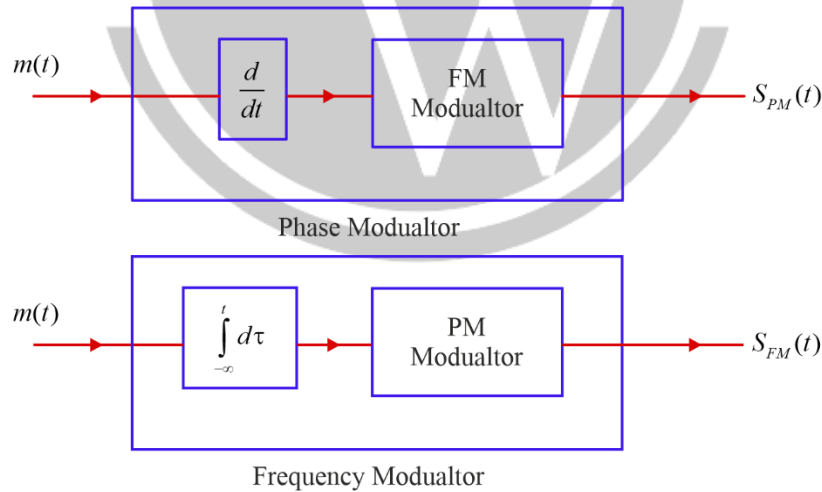
### Frequency Mixture and Multiplier

Mixture/Multiplier Input	Mixture Output	(Multiplied by $n$ ) Multiplier Output
$A_c$	$A'_c$	$A'_c$
$f_c$	$ f_c - f_L $ or $f_c + f_L = f'_c$	$nf_c$
$\beta$	$\beta$	$n\beta$
$f_m$	$f_m$	$f_m$
$\Delta f$	$\Delta f$	$n\Delta f$
BW	BW	$(n\beta + 1)2f_m$
Spectral spacing	$f_m$	$f_m$
Frequency components	$f'_c, f'_c \pm f_m, f'_c \pm 2f_m$	$nf_c, nf_c \pm f_m, nf_c \pm 2f_m$

### Wideband Angle Modulation generation –

$$PM[m(t)] = FM\left[\frac{dm(t)}{dt}\right] \text{ If } K_p = K_f = K$$

$$FM[m(t)] = PM\left[\int_{-\infty}^t m(\tau) d\tau\right]$$



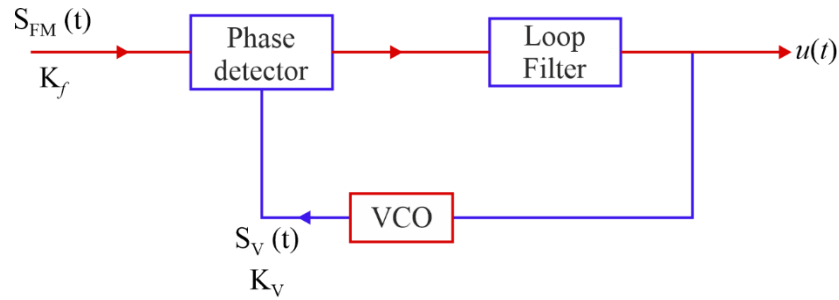
### Wideband FM Generation Methods

1. Armstrong Method (Indirect Method)
2. Direct Method
  - VCO (Voltage Controlled Oscillator) is used. It is modified version of Hartley oscillator

$$\frac{\Delta\omega}{\omega_c} = \frac{\Delta C}{2C_0}$$

### FM Demodulator

1. Theoretical method
2. Practical method. PLL (Phase Locked Loop)



$$(1) \quad v(t) = \frac{K_f}{K_v} m(t)$$

- (2) Lock mode  $\rightarrow$  Frequency lock  
Capture mode  $\rightarrow$  Phase lock
- (3)  $L.R \geq C.R$

### Super Hetrodyne Receiver

$f_L$  = Local oscillator frequency

$f_s$  = Desired frequency

$f_{Si}$  = Frequency of image station

**Case 1 :** If relation between  $f_l$  and  $f_s$  is not mentioned.

Assume :  $f_l > f_s$

1.  $f_l = f_s + IF$
2.  $f_{Si} = f_l + IF$
3.  $f_{Si} = f_s + 2IF$

**Case 2 :** When relation between  $f_i$  and  $f_s$  is given

If  $f_{Si} < f_l < f_s$       If  $f_s < f_s < f_{sl}$ , then Case 1

- then
1.  $f_s = f_l + IF$
  2.  $f_l = f_{Si} + IF$
  3.  $f_s = f_{Si} + 2IF$

### Image Rejection Ratio

$$IRR = \sqrt{1 + P^2 Q^2}$$

$Q$  : Quality factor of Oscillator

$$P = \frac{f_{Si}^2 - f_s^2}{f_{Si} f_s} \quad f_{Si} > f_s \quad P^2 Q^2 \gg 1$$

$$IRR = PQ$$



# 3

## RANDOM VARIABLE AND RANDOM PROCESS

### 3.1. Introduction

#### Random variable → Real and complex

- R.V. is a function performing mapping from sample space of R.E. to real line.
- $X(\lambda)$ : Random variable
- Domain of R.V.  $\rightarrow \lambda$  (Sample point)
- Range of R.V.  $\rightarrow$  Subset of real line
- One to one or many to one mapping
- $P\{X \leq a\} \rightarrow$  Probability of set in which all the comes satisfy  $x(\lambda) \leq a$ .

#### CDF of R.V.

Let random variable X,  $x \rightarrow$  Values taken by R.V.

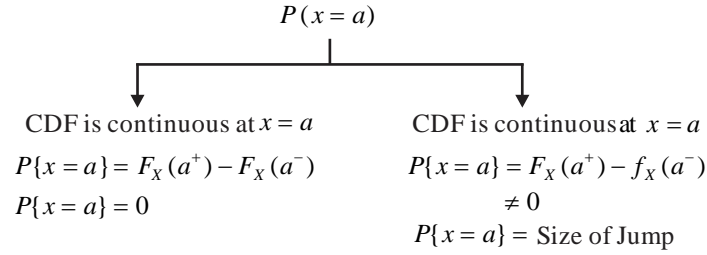
- (1)  $F_X(x) = P\{X \leq x\} = 1 - P\{X > x\}$
- (2)  $F_X(a) = P\{X \leq a\} = 1 - P\{X > a\}$
- (3)  $F_{|X|}(y) = P\{|X| \leq y\} = P\{-y \leq X \leq y\}$

#### Properties

- (1)  $F_X(\infty) = 1$
- (2)  $F_X(-\infty) = 0$
- (3)  $F_X(\infty) + F_X(-\infty) = 1$
- (4)  $F_X(x) = P\{X \leq x\} \Rightarrow 0 \leq F_X(x) \leq 1$ 
  - (a) CDF always non negative.
  - (b) Lower bound:  $F_X(x) = 0$ , upper Bound = 1
- (5) CDF is monotonically non decreasing function of  $x \left( \frac{dF_X(x)}{dx} \geq 0 \right)$
- (6) Graph of CDF is always amplitudes continuous from right.

#### • Key point :

- (1)  $P\{a < x \leq b\} = F_X(b^+) - F_X(a^+)$
- (2)  $P\{a \leq x \leq b\} = F_X(b^+) - F_X(a^-)$
- (3)  $P\{a < x < b\} = F_X(b^-) - F_X(a^+)$
- (4)  $P\{a \leq x < b\} = F_X(b^-) - F_X(a^-)$



### Probability Density Function

Random variable  $X$

$x \rightarrow$  Variable taken by R.V.

$f_X(x) \rightarrow$  Symbol

$$f_X(x) = \frac{dF_X(x)}{dx}$$

$$f_X(x) = \int_{-\infty}^x f_X(x) dX$$

$$f_X(x) = \int_{-\infty}^a f_X(\lambda) d\lambda$$

### Properties :

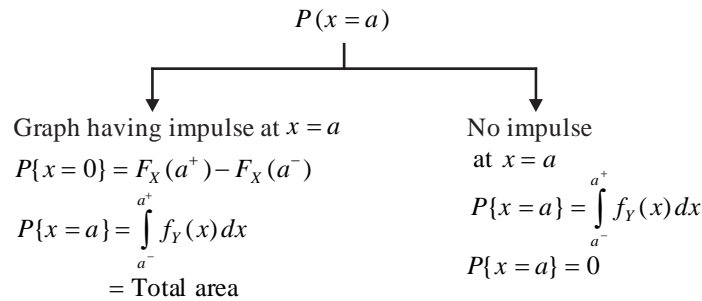
- (1)  $f_X(x) \geq 0 \rightarrow$  Non negative
- (2)  $0 \leq f_X(x) < \infty \rightarrow$  Upper bound  

↓

Lower bound
- (3)  $F_X(\infty) = \int_{-\infty}^{\infty} f_X(x) dx = 1$
- (4) Graph of PDF can be even or NENO but cannot be odd.
- (5)  $P\{-\infty < X \leq x\} = \int_{-\infty}^x f_X(\lambda) d\lambda$
- (6)  $P\{a < X \leq b\} = \int_{a^+}^{b^+} f_X(x) dx$

$$(7) \quad P\{a \leq X \leq b\} = \int_{a^-}^{b^+} f_X(x) dx$$

$$(8) \quad P\{a < X < b\} = \int_{a^+}^{b^-} f_X(x) dx$$



### Discrete Random Variable:

- (1) PDF should have impulses only.
- (2) CDF should have staircase only.

(1) **Probability mass function of DRV** : Let  $X$  is D.R.V.

$P_X(x) = P(X=x)$  probability such that  $X=x$

- $0 \leq P_X(x) \leq 1$
- $\sum_x P_X(x) = 1$

(2) **PDF of a D.R.V** : Let  $X$  is O.R.V.

$$f_X(x) = \sum_i P_X(x_i) \delta(x-x_i) = \sum_i P(x=x_i) \delta(x-x_i)$$

(3) **CDF of a D.R.V.** : Let  $X$  is D.R.V.

$$F_X(x) = \int_{-\infty}^x f_X(x) dx$$

$$F_X(x) = \sum_x P\{X=x_i\} u\{x-x_i\}$$

(4)  $P\{X=a\}$  may or may not be zero.

### Continuous Random Variable

Maps sample point to continuous range of values on real axis.

(1) PDF of C.R.V should not contain impulses at all.

(2) CDF of C.R.V

- Should not contain jump type discontinuity
- It should be amplitude continuous every where

(3) PMF not defined for C.R.V because for CRV  $P\{X=a\}$  will always be zero.



$$P(A/B) = \frac{P(A \cap B)}{P(B)} \rightarrow \text{Conditional probability of A given B.}$$

$$P(A \cap B) = P(B)P(A/B) = P(A)P\left(\frac{B}{A}\right) = \text{Joint probability.}$$

**Expectation operator** : Performs operations on R.V. only.

### Linear Operator

$$E[C] = C, E[C^2] = C^2 \quad E(X) = \begin{cases} \int_{-\infty}^{\infty} xf_X(x)dx & X : CRV \\ \sum_i x_i P\{X = x_i\} & X : DRV \end{cases}$$

$$E[aX] = aE[X]$$

$$E[aX + b] = aE[X] + E[b]$$

$$E[ag(X) + bH(y)] = aE[g(x)] + bE[H(y)]$$

$$E[X^2] = \int_{-\infty}^{\infty} x^2 f_X(x) dx$$

$$E[g(x)] = \int_{-\infty}^{\infty} g(x) f_X(x) dx$$

### Gaussian Random Variable

CRV X is having Gaussian or random distribution.

X is having Gaussian PDF, X is called G.R.V.

$$E[X] = \mu_X, E[(X - \mu_X)^2] = \text{Variance} = \sigma_X^2$$

$$X \sim N\{\mu_X, \sigma_X^2\} \quad f_X(x) = \frac{1}{\sqrt{2\pi\sigma_X^2}} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}} \quad -\infty < x < \infty$$

### Key Point :

$$(1) \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma_X^2}} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}} dx = 1$$

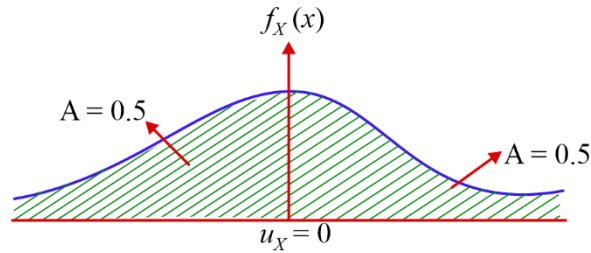
$$(2) \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\sigma_X^2}} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}} dx = \mu_X = E[X]$$

$$(3) \int_{\mu_X}^{\infty} \frac{1}{\sqrt{2\pi\sigma_X^2}} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}} dx = \int_{-\infty}^{\mu_X} \frac{1}{\sqrt{2\pi\sigma_X^2}} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}} dx = \frac{1}{2}$$

### Zero mean Gaussian distribution-

$$X \sim N(\mu_X, \sigma_X^2) \Rightarrow X \sim N[0, \sigma_X^2] \Rightarrow E[X] = 0$$

$$f_X(x) = \frac{1}{\sqrt{2\pi\sigma_X^2}} e^{-\frac{x^2}{2\sigma_X^2}}$$



Zero Mean, unit variance :

$$X \sim N(0,1) \quad f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}, \quad \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = 1$$

Q- function :

$$Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^{\infty} e^{-z^2/2} dz \quad \text{as } x \uparrow, Q(x) \downarrow$$

$$Q(\infty) = 0, \quad Q(-\infty) = 1, \quad Q(0) = 0.5, \quad Q(x) + Q(-x) = 1$$

$$P[X > z] = Q(P) = Q\left[\frac{z - \mu_X}{\sigma_X}\right]$$

$$f_X(z) = P(X \leq z) = 1 - P(X > z) = 1 - Q\left[\frac{z - \mu_X}{\sigma_X}\right]$$

### Statistical averages of a R.V.

$n^{\text{th}}$  order moment about origin-

$$E[(X - 0)^n] = E[X^n] = \begin{cases} \int_{-\infty}^{\infty} x^n f_X(x) dx & X : \text{CRV} \\ \sum_i x_i^n P\{X = x_i\} & X : \text{DRV} \end{cases}$$

### 1<sup>st</sup> order moment about origin

$$E[X] = \int_{-\infty}^{\infty} x f_X(x) dx = \sum_i x_i P\{X = x_i\}$$

$$E[X] = \bar{X} = \mu_X = m_1 \rightarrow \text{dc value, avg. value Mean value}$$

$$[E[X]]^2 \rightarrow \text{d.c. power}$$

## 2<sup>nd</sup> order moment about origin-

$$E[(X-0)^2] = E[X^2] = \bar{X}^2 \begin{cases} \int_{-\infty}^{\infty} x^2 f_X(x) & X : CRV \\ \sum_i x_i P\{X = x_i\} & X : DRV \end{cases}$$

$E[X^2]$  = Mean square value of R.V.  $X$  = Total power of R.V.  $x$

1<sup>st</sup> order moment about mean -  $E[(X - \mu_X)] = 0$

2<sup>nd</sup> order moment about mean -  $E[(X - \mu_X)^2] = E[X^2] - \mu_X^2$

$$\sigma_X^2 = E[X^2] - \mu_X^2$$

$\downarrow$   
A.C.  
Power

$\downarrow$   
Total  
Power

$\downarrow$   
dc  
Power

### Important point:

(1)  $\sigma_X^2 \geq 0, E[X^2] \geq \mu_X^2$

(2) If  $X$  is zero mean R.V.

$$E[X^2] = \sigma_X^2, \text{MSV}(X) = \text{Var}(X)$$

(3) Standard deviation

$$\sqrt{\text{Variance}} = \sqrt{\sigma_X^2} = \pm \sigma_X$$

(4)  $Y = aX + b$

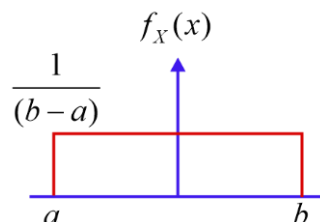
$$E[Y^2] = a^2 E[X^2] + b^2 + 2ab E[X]$$

$$\sigma_Y^2 = a^2 \sigma_X^2$$

## Standard Distribution of R.V.

(1) Uniform distribution  $X \sim U[a, b]$

$$f_X(x) = \begin{cases} \frac{1}{(b-a)} & a \leq X \leq b \\ 0 & \text{otherwise} \end{cases}$$

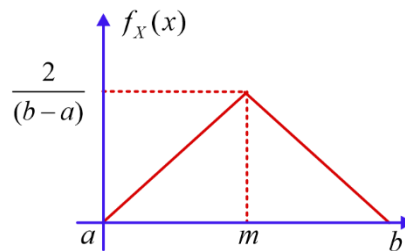


$$E[X] = \frac{a+b}{2}, \quad E[X^2] = \frac{a^2 + b^2 + ab}{3}, \quad \sigma_X^2 = \frac{(b-a)^2}{12}$$

(2) Triangular distribution

$$X \sim \text{tri}(a, m, b)$$

$$E[X] = \frac{a+m+b}{3}$$



(3) Rayleigh Distribution

$$X \rightarrow \text{CRV}$$

$$f_X(x) = \begin{cases} \frac{x}{\sigma_X^2} e^{\frac{-x^2}{2\sigma_X^2}} & x \geq 0 \\ 0 & \text{else} \end{cases}$$

$$\int_0^{\infty} \frac{x}{\sigma_X^2} e^{\frac{-x^2}{2\sigma_X^2}} dx = 1$$

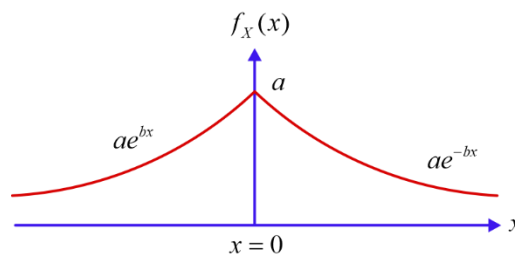
If X and Y are two G.R.V. Then  $Z = \sqrt{X^2 + Y^2}$  will have reyleigh distribution.

(4) **Exponential Distribution** : If CRV has exponential distribution then it will have PDF

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x < 0 \end{cases} \quad \int_0^{\infty} \lambda e^{-\lambda x} dx = 1$$

Laplacian Distribution

$$X \rightarrow \text{CRV}$$



$$f_X(x) = ae^{-b|x|} \quad -\infty < x < \infty$$

$$\text{If } \frac{2a}{b} = 1, \quad a > 0, b > 0$$

$$f_X(x) = \begin{cases} ae^{bx} & x < 0 \\ ae^{-bx} & x > 0 \end{cases}$$

### Discrete Random variable-Binomial, Position distribution

#### Binomial distribution necessary condition:-

- (1) The no of trials  $n$  should be finite.
- (2) Trials are independent
- (3) Each trials should result in 2 outcomes success or failure.
- (4) Prob of success in each trial should be constant.

PMF:

$$P\{X = r \text{ success}\} = n_c p_r q^{n-r}$$

$$E[X] = \sum_i x_i p\{X = x_i\} = n_p \quad \sigma_X^2 = npq$$

$$E[X^2] = npq + (np)^2$$

$$\text{Std deviation } \sigma_X = \pm \sqrt{npq}$$

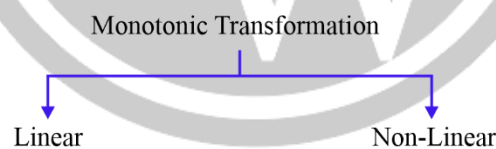
#### Position Distribution

Specific type of binomial distribution where  $n \rightarrow \infty$

$n \rightarrow$  very large,  $p \rightarrow$  very small,  $np \rightarrow$  finite  $\lambda = np$

$$p\{X = r\} = \frac{\lambda^r e^{-\lambda}}{r!} \text{ probability of } X = r \text{ (success)}$$

$$E[X] = \lambda, \sigma_X^2 = \lambda$$



If  $Y = g(X)$  is having monotonic  $T_X$ .

Given  $X \xrightarrow{PDF} f_X(x)$ ,

$$f_Y[y] = \left\{ f_X[x] \left| \frac{dx}{dy} \right| \right\} \text{ function of } y$$

$$Y = aX + b \begin{cases} \rightarrow \text{X:CDF } f_X(x) \\ \rightarrow \text{X:PDF } f_X(x) \end{cases}$$

(1) case  $a > 0$

$$F_Y[y] = F_X\left(\frac{y-b}{a}\right), f_Y(y) = \frac{1}{a} f_X\left(\frac{y-b}{a}\right)$$

(2)  $Y = -aX + b \quad a > 0$

$$F_Y(y) = 1 - F_X\left(\frac{y-b}{-a}\right), f_Y(y) = \frac{1}{a} f_X\left(\frac{y-b}{-a}\right)$$

### Monotonic linear $T_X$ :

$$y = aX + b$$

$$X \sim U[m_1, m_2] \rightarrow Y \sim U[am_1 + b, am_2 + b]$$

$$X \sim \Delta[m_1, m_2, m_3] \rightarrow Y \sim \Delta[am_1 + b, am_2 + b, am_3 + b]$$

$$X \sim N[\mu_X, \sigma_X^2] \rightarrow Y \sim N[\mu_Y, \sigma_Y^2]$$

$$Y \sim N[a\mu_X + b, a^2\sigma_X^2]$$

### Monotonic Non-Linear $T_X$ :

$$X \rightarrow f_X(x)$$

$$Y \rightarrow X^3, f_Y(y) = ?$$

$$f_Y(y) = \left\{ f_X(x) \left| \frac{dx}{dy} \right| \right\}$$

$$f_Y(y) = \frac{1}{3y^{2/3}} f_X(y^{1/3})$$

### Non - Monotonic $T_X$ :

$$Y = y, g(X) = y, X = g^{-1}(y)$$

$$\begin{cases} \rightarrow x_1 \\ \rightarrow x_2 \\ \rightarrow x_3 \end{cases}$$

$$f_Y(y) = f_X(x_1) \left| \frac{dx_1}{dy} \right| + f_X(x_2) \left| \frac{dx_2}{dy} \right| + \dots$$

### 2D Random variable :

$$(X, Y) \rightarrow 2\text{DR.V.} \begin{cases} \rightarrow F_{X,Y}(x, y) = \text{Joint CDF} \\ \rightarrow f_{X,Y}(x, y) = \text{Joint PDF} \\ \rightarrow P_{XY}(x_i, y_i) = \text{Joint PMF} \end{cases}$$

If A and B are independent

$$P\left(\frac{A}{B}\right) = P(A), \quad P\left(\frac{B}{A}\right) = P(B), \quad P(A \cap B) = P(A)P(B)$$

$$F_{X,Y}(x,y) = F_X(x)F_Y\left(\frac{y}{x}\right) = F_Y(y)F_X\left(\frac{x}{y}\right)$$

Diagram illustrating the relationship between the joint CDF and marginal/conditional CDFs:

- $F_X(x)$  is labeled as the **Marginal CDF of X**.
- $F_Y(y)$  is labeled as the **Marginal CDF of Y**.
- $F_Y\left(\frac{y}{x}\right)$  is labeled as the **Conditional CDF of Y given X**.
- $F_X\left(\frac{x}{y}\right)$  is labeled as the **Conditional CDF of X given Y**.

### If X and Y are independent R.V.

$$F_{XY}(x,y) = f_X(x)f_Y(y)$$

$$\Rightarrow f_{XY}(x,y) = f_X(x)f_Y\left(\frac{y}{x}\right) = f_Y(y)f_X\left(\frac{x}{y}\right)$$

If X and Y are independent R.V.

$$f_{XY}(x,y) = f_X(x)f_Y(y)$$

$$\Rightarrow P_{XY}(x_i, y_j) = P_X(x_i)P_Y\left(\frac{y_j}{x_i}\right) = P_Y(y_j)P_X\left(\frac{x_i}{y_j}\right)$$

If X and Y are independent R.V.

$$P_{XY}(x_i, y_j) = P_X(x_i)P_Y(y_j)$$

Joint CDF = Let (X,Y) are BIVARIATE R.V.

$$F_{XY}(x,y) = P\{X \leq x \cap (Y \leq y)\} = P\{X \leq x; Y \leq y\}$$

### Properties:

- (1)  $0 \leq F_{XY}(x,y) \leq 1$
- (2)  $F_{XY}(-\infty, y) = P\{(X \leq -\infty) \cap (Y \leq y)\} = 0$
- (3)  $F_{XY}(x, -\infty) = 0$
- (4)  $F_{XY}(-\infty, -\infty) = 0$
- (5)  $F_{XY}(\infty, \infty) = 1$
- (6)  $F_{XY}(x_1, y_1) = P\{(X \leq x_1) \cap (Y \leq y_1)\}$
- (7)  $P\{(x_1 < X \leq x_2) \cap (y_1 < Y \leq y_2)\}$   
 $= F_{XY}(x_1^+, y_1^+) + F_{XY}(x_2^+, y_2^+) - F_{XY}(x_1^+, y_2^+) - F_{XY}(x_2^+, y_1^+)$
- (8)  $F_{XY}(x,y) = F_X(x)F_Y\left(\frac{y}{x}\right) = F_Y(y)F_X\left(\frac{x}{y}\right)$
- (9) X and Y are independent R.V.  
 $F_{XY}(x,y) = F_X(x)F_Y(y)$
- (10)  $F_X(x,y) = F_{XY}(x,\infty), \quad F_Y(y) = F_{XY}(\infty,y)$

### Conditional CDF

$$F_{\frac{X}{Y}}\left(\frac{x}{y}\right) = \frac{F_{XY}(x, y)}{F_Y(y)} \quad \text{of} \quad F_Y(y) \neq 0$$

$$F_{\frac{X}{Y}}\left(\frac{x}{y}\right) = \frac{P[(X \leq x) \cap (Y \leq y)]}{P[(X \leq \infty) \cap (Y \leq y)]}$$

### Joint PDF

$$f_{XY}(x, y) = \frac{\partial^2 F_{XY}(x, y)}{\partial x \partial y}$$

$$F_{XY}(x, y) = \int_{-\infty}^x \int_{-\infty}^y f_{XY}(u, v) du dv$$

$$F_{XY}(x, y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy = 1$$

### Marginal PDF

$$(1) \quad f_X(x) = \int_{-\infty}^{\infty} f_{XY}(x, y) dy, \quad f_Y(y) = \int_{-\infty}^{\infty} f_{XY}(x, y) dx$$

If X and Y are independent  $f_{XY}(x, y) = f_X(x)f_Y(y)$

$$f_{XY}(x, y) = f_X(x)f_Y\left(\frac{y}{x}\right) = f_Y(y)f_X\left(\frac{x}{y}\right)$$


### Conditional PDF

$$f_{XY}(x, y) = \frac{f_{XY}(x, y)}{f_Y(y)} = \frac{\int_{-\infty}^x \int_{-\infty}^y f_{XY}(x, y) dx dy}{\int_{-\infty}^{\infty} f_{XY}(x, y) dx}$$

### Probability Calculation in 2-D region

#### Given Joint PDF

$$f_{XY}(x, y) = \begin{cases} \text{define} & [(x_1 < X \leq x_2) \cap (y_1 < Y \leq y_2)] \\ 0 & \text{else} \end{cases}$$


**R<sub>2</sub>**  
 Region in which PDF is defined

$$P\{(a < X \leq b) \cap (c < y \leq d)\} = ?$$

**R<sub>1</sub>** : Region in which probability has to be calculated.

#### Method:



$$P(X, Y \in R_1) = \iint_R f_{XY}(x, y) dx dy \quad (R = R_1 \cap R_2)$$

(1) X and Y not independent R.V.

$$P(X, Y \in R_1) = \iint_{R_1} f_{X,Y}(x) f_Y(Y) dx dy \quad R = (R_1 \cap R_2)$$

(Central Limit Theorem)

If X and Y are D.R.V

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

$$P_{XY}(x_i, y_j) = P\{(X = x_i) \cap (Y = y_j)\}$$

Joint PMF

**Marginal PMF :**

$$P_X(x_i) = \sum_j P_{XY}(x_i, y_j)$$

$$P_Y(y_j) = \sum_i P_{XY}(x_i, y_j)$$

**Minimum of 2 independent R.V.**

X, Y are two I.R.V

$$\min(X, Y) > Z = (X > Z) \cap (Y > Z)$$

$$P[\min(X, Y) > Z] = P[X > Z]P[Y > Z] = \iint_R f_{XY}(x, y) dx dy$$

$$P[\min(X, Y) \leq Z] = 1 - P[\min(X, Y) > Z]$$

$$\underbrace{P[\min(X, Y) > Z]}_R = \iint_R f_X(x) f_Y(y) dx dy$$

Let  $Z = \max(X, Y) \rightarrow$  R.V.

$$\text{CDF of } Z \quad F_Z(Z) = F_X(Z) \cdot F_Y(Z)$$

$$\text{PDF of } Z \quad f_Z(Z) = F_X(Z) f_Y(Z) + F_Y(Z) f_X(Z)$$

Let  $Z = \min[X, Y] \rightarrow$  R.V.

$$\text{CDF of } Z \quad F_Z(Z) = f_X(Z) + g_Y(Z) + f_Y(Z) F_X(Z)$$

$$\text{PDF of } Z \quad f_Z(Z) = f_X(Z) + f_Y(Z) - F_X(Z) f_Y(Z) - F_Y(Z) f_X(Z)$$

**Statistical parameters of 2D R.V.**

(1)  $(k, r)^{th}$  order joint moment about origin  $E[X^k Y^r]$   $(1, 1)^{st}$  order joint moment about origin.

$$E[X^1, Y^1] = E[XY] = R_{XY} \rightarrow \text{Cross correlation between R.V. X and Y.}$$

➤  $E[XY] = R_{XY} = 0 \rightarrow$  R.V. X and Y are orthogonal.

(2)  $(k, r)^{th}$  order joint moment about mean-

$$E[(X - \bar{X})^k (Y - \bar{Y})^r]$$

(1,1)<sup>st</sup> order joint Moment about mean-

$$E[(X - \bar{X})(Y - \bar{Y})] = E[XY] - \bar{X}\bar{Y} = \text{cov}(X, Y)$$

$$\text{cov}(X, Y) = \sigma_{XY} = E[XY] - E[X]E[Y] = R_{XY} - \mu_x \mu_y$$

When 2 R.V. X and Y are uncorrelated-

$$\text{cov}(X, Y) = 0, \quad E[XY] = E[X]E[Y]$$

➤  $E[X^k Y^r] = E[X^k]E[Y^r]$  X, Y are independent.

➤ If 2 R.V. are independent then they has to be uncorrelated but converse is not necessarily true.

### One function of two R.V.

$$W = aX + bY$$

$$(1) \quad E[W] = aE[X] + bE[Y]$$

$$(2) \quad E[W^2] = a^2 E[X^2] + b^2 E[Y^2] + 2ab R_{XY}$$

$$(3) \quad \sigma_W^2 = a^2 \sigma_X^2 + b^2 \sigma_Y^2 + 2ab \text{cov}(X, Y)$$

One function of Three R.V.  $W = aX_1 + bX_2 + cX_3$

$$(1) \quad E[W] = a\mu_{X_1} + b\mu_{X_2} + c\mu_{X_3}$$

$$(2) \quad E[W^2] = a^2 X_1^2 + b^2 X_2^2 + c^2 X_3^2 + 2abX_1X_2 + 2bcX_2X_3 + 2caX_1X_3$$

$$(3) \quad \sigma_W^2 = a^2 \sigma_{X_1}^2 + b^2 \sigma_{X_2}^2 + c^2 \sigma_{X_3}^2 + 2ab \text{cov}(X_1, X_2) + 2bc \text{cov}(X_2, X_3) + 2ca \text{cov}(X_1, X_3)$$

$$\text{Var}(X + Y) = \text{Var}(X - Y)$$

Only when X, Y are  $\rightarrow$  uncorrelated and independent

### Correlation coefficient

$$\rho(X, Y) = \frac{\sigma_{XY}}{\sigma_X \sigma_Y} = \frac{\text{cov}(X, Y)}{(\text{Std.dev.of } X) \times (\text{std.dev.of } Y)}$$

$$\text{➤ } -1 \leq \rho \leq 1$$

$$\text{➤ } \rho(X, X) = 1, \quad \rho(X, -X) = -1$$

$$\text{➤ } X, Y \text{ are independent } \rho(X, Y) = 0$$

Let X, Y are two R.V.

$$E\left[\frac{g(Y)}{X=x}\right] = \int_{-\infty}^{\infty} g(y) f_{\frac{Y}{X}}\left(\frac{y}{x}\right) dy$$

$$E\left[\frac{g(X)}{Y=y}\right] = \int_{-\infty}^{\infty} g(x) f_{\frac{X}{Y}}\left(\frac{x}{y}\right) dx$$

### Calculation of probability in n-D region

#### Theorem -1

If  $X_1, X_2, X_3, \dots, X_n$  are statistically independent random variables.

Let  $Z = X_1 + X_2 + \dots + X_n$

$$\begin{matrix} \uparrow & \uparrow & \uparrow \\ f_{X_1}(z) & f_{X_2}(z) & f_{X_n}(z) \end{matrix}$$

$$f_Z(z) = f_{X_1}(z) * f_{X_2}(z) * \dots * f_{X_n}(z)$$

When and only when all the R.V. are statistically independent.

➤ R.V. are linearly combined.

#### Theorem-2

$X_1, X_2, X_3, \dots, X_n$  are statically independent non Gaussian R.V.

$$Z = X_1 + X_2 + X_3 + \dots + X_n$$

$$f_Z(z) = f_{X_1}(z) * f_{X_2}(z) * \dots * f_{X_n}(z)$$

If  $n \rightarrow \infty$   $f_Z(z)$  = Gaussian irrespective of nature of  $(X_i)_{i=1}^n$

#### Theorem-3

$X_1, X_2, X_3, \dots, X_n$  are statistically independent G.R.V.

$$Z = X_1 + X_2 + \dots + X_n$$

$$f_Z(z) = f_{X_1}(z) + f_{X_2}(z) + \dots + f_{X_n}(z)$$

$n \rightarrow \text{Finite} | \text{infinite}, \rightarrow Z: \text{GRV}$

### Problem Solving Technique :

**Case 1 :**  $X_1, X_2, \dots, X_n$  are statistically independent G.R.V.

$$P[X_1 + X_2 + X_3 > a] = P(Z > a) = \int_a^{\infty} f_Z(z) dz = 1 - \int_{-\infty}^a f_Z(z) dz$$

$\downarrow$   
 Non GRV

Where  $Z = X_1 + X_2 + X_3$

$$f_Z(z) = f_{X_1}(z) \times f_{X_2}(z) \times f_{X_3}(z)$$

**Case 2 :** If  $X_1, X_2, X_3$  are statistically independent G.R.V.

$$P(X_1 + X_2 + X_3 > a) = P[Z > a] = Q\left[\frac{a - \mu_z}{\sigma_z}\right]$$

$$Z = X_1 + X_2 + X_3$$

$$\mu_z = \mu_{X_1} + \mu_{X_2} + \mu_{X_3}, \quad \sigma_z^2 = \sigma_{X_1}^2 + \sigma_{X_2}^2 + \sigma_{X_3}^2$$

**Note :** If  $X_1, X_2, X_3, \dots, X_n$  are I.I.D. random variables

$$P(\text{one of them is largest}) = \frac{1}{n}$$

$$P(\text{one of them is smallest}) = \frac{1}{n}$$

### Random Process

$$X(\lambda, t) = \{X(\lambda_1, t), X(\lambda_2, t)\} \rightarrow \begin{array}{l} \text{Collection of sample function} \\ \text{of} \\ \text{Euemle of sample function} \end{array}$$

Random process or Random signal  
or stochastic signal

$X(\lambda_1, t_1) \rightarrow$  sample value, values taken by R.V. When R.P. is observed at  $t = t_1$

C.T.R.P  $\rightarrow$  It maps the sample points onto continuous time sample function, collection of continuous time sample function.

$$X(t) = A \cos(\omega_0 t + \phi) \xrightarrow{t=t_1} X(t_1) = A \cos(\omega_0 t_1 + \phi)$$

C.T.R.P R.V., D.R.V.

$t \rightarrow$  Continuous time,  $A \rightarrow$  Constant,  $\omega_0 \rightarrow$  constant

$\phi \sim U[-\pi, \pi] \rightarrow$  CRV

Any typical R.P can be understood as  $x(t) = f(t, \phi)$

(Function of time and R.V.)  $x(n) = f(n, \phi)$

### Statistical parameter of R.P.

**Case 1 :**  $X(t) \xrightarrow{t=t_0} X(t_0) - \text{CRV}$

$$E[X(t_0)] = \int_{-\infty}^{\infty} x f_{X(t_0)}(x) dx$$

$$E[x^2(t_0)] = \int_{-\infty}^{\infty} x^2 f_{X(t_0)}(x) dx, \quad \sigma_{X(t_0)}^2 = E[X^2(t_0)] - 9E(X(t_0))^2$$

**Case 2 :**  $X(t) \xrightarrow{t=t_0} X(t_0) - \text{DRV}$   
CTRP

$$E[X(t_0)] = \sum_i x_i P_{X(t_0)}(x_i) = \sum_i x_i P\{X(t_0) = x_i\}$$

$$E[x^2(t_0)] = \sum_i x_i^2 P_{X(t_0)}(x_i) = \sum_i x_i^2 P\{X(t_0) = x_i\}$$

$$\sigma_{X(t_0)}^2 = E[X^2(t_0)] - (E[X(t_0)])^2$$

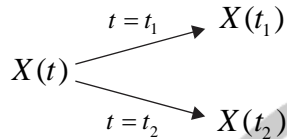
**Case 3 :** DTRP  $\rightarrow$  CRV

$$E[X(n_0)], E[x^2(n_0)], \sigma_{X(n_0)}^2 \rightarrow \text{Some as case 1, replace to by } n_0$$

**Case 4 :** DTRP  $\rightarrow$  DRV

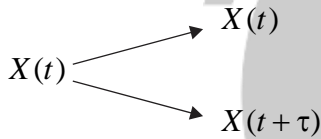
$$E[X(n_0)], E[x^2(n_0)], \sigma_{X(n_0)}^2 \rightarrow \text{Same as case-2, Replace to by } n_0$$

### CTR.P



$$E[X(t_1)X(t_2)] = R_{X(t_1)X(t_2)} = R_{XX}(t_1, t_2)$$

Auto correlation of RP  $X(t)$



$$\text{Then } E[X(t_1)X(t+\tau)] = R_{XX}(t, t+\tau)$$

$$\text{Cov}(X(t_1)X(t_2)) = E[X(t_1)X(t_2)] - E[X(t_1)]E[X(t_2)]$$

$$\sigma_{XX}(t_1, t_2) = R_{XX}(t_1, t_2) - \mu_{X(t_1)}\mu_{X(t_2)}$$

Auto covariance of R.P.  $X(t)$

$$\sigma_{XX}(t, t+\tau) = R_{XX}(t, t+\tau) - \mu_{X(t)}\mu_{X(t+\tau)}$$

### Cross Correlation

$$\begin{matrix} X(t) & \xrightarrow{t=t_1} & X(t_1) & , & Y(t) & \xrightarrow{t=t_2} & Y(t_2) \\ R.P & & R.V & & R.P & & R.V \end{matrix}$$

$$E[X(t_1)Y(t_2)] = R_{XY}(t_1, t_2)$$

(1) If  $R_{XY}(t_1, t_2) = 0 \forall t_1 \in \text{TR}$   $X(t)$  and  $Y(t)$  R.P. will

$t_2 \in \text{TR}$  Become orthogonal.

(2)  $\text{Cov}[X(t_1), Y(t_2)] = R_{XY}(t_1, t_2) - \mu_{X(t_1)}\mu_{X(t_2)}$

$$= 0 \quad \forall \quad t_1 \in \text{TR}$$

$$t_2 \in \text{TR}$$

RP  $X(t)$  and  $Y(t)$  are uncorrelated.

If  $X(t_1)$  and  $X(t_2)$  are independent

$$E[X(t_1)X(t_2)] = \begin{cases} E[X(t_1)E[X(t_2)]] & t_1 \neq t_2 \\ E[X^2(t_1)] & t_1 = t_2 \end{cases}$$

Same for DRV, replace  $t$  by  $n$ .

### Types of R.P.

(1) Strict sense stationary R.P.  $\rightarrow$  R.P. should be independent of time shift

$$X(t) \rightarrow \frac{X(t_1)X(t_2) \dots X(t_k)}{kRV.}$$

### $K^{\text{th}}$ order Joint PDF-

$$X(t): \frac{(x_1, x_2 \dots x_k) f_X(t_1)X(t_2) \dots X(t_k) X(t_1 + \tau), X(t_2 + \tau) \dots X(t_k + \tau)}{k \text{ R.V.}} \dots(i)$$

### $K^{\text{th}}$ order Joint PDF-

$$(x_1, x_2 \dots x_k) f_{X(t_1+\tau)X(t_2+\tau) \dots X(t_k+\tau)} \dots(ii)$$

(i) = (ii)  $\rightarrow X(t)$  is solid to be SSSRP.

$$f_{X(t_1)}(x) = f_{X(t_2+\tau)}(x) \text{ independent of time}$$

2<sup>nd</sup> order joint PDF is independent of time shift.

$$f_{X(t_1)X(t_2)}(x_1, x_2) \rightarrow f_{X(0)X(t_2-t_1)}(x_1, x_2)$$

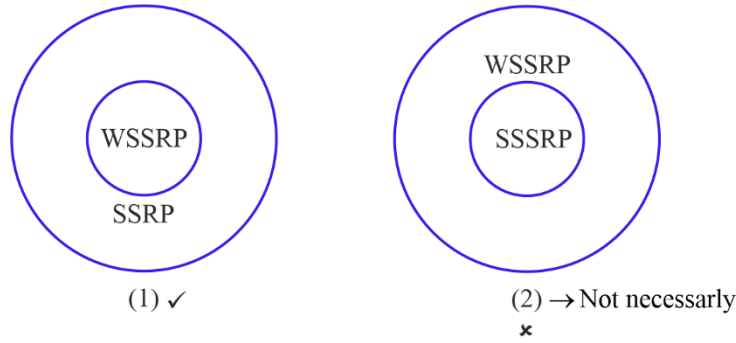
- Does not depend on individual sampling instances  $t_1$  and  $t_2$
- Depends on time difference between sampling instances  $t_1$  and  $t_2$

$$E[X(t_1)X(t_2)] = E[X(0)X(t_2 - t_1)] = R_{XX}(t_1, t_2)$$

$$E[X(0)X(t_2 - t_1)] = R_{XX}(0, t_1 - t_2) = R_{XX}(t_1 \sim t_2)$$

$$\sigma_{XX}(t_1 \sim t_2) = R_{XX}(t_1 \sim t_2) - \mu_X^2$$

WSSRP  $\rightarrow$  There are stationary RP which are stationary at least upto 2<sup>nd</sup> order.



(1)  $E[X(t)] = \mu_X$  Constant

(2)  $E[X^2(t)] = \text{Constant}$

(3)  $\sigma_{X(t)}^2 = \text{Constant}$

$$E[RP] = E[RV]$$

$$\text{MSV}(RP) = \text{MSV}(RV)$$

$$\text{Var}(RP) = \text{Var}(RV)$$

(4)  $E[X(t_1)X(t_2)] = R_{XX}(t_1 \sim t_2)$

$$E[X(t+\tau)X(t)] = R_{XX}(-\tau)$$

$$E[X(t)X(t+\tau)] = R_{XX}(\tau)$$

$$R_{X_X}(\tau) = R_{XX}(-\tau)$$

$$\sigma_{X(t)}^2 = R_{XX}(0) - \mu_X^2$$

$$\text{Cov}[X(t)X(t+\tau)] = R_{X_X}(\tau) - \mu_X^2$$

(5)  $E[X^2(t)] = R_{XX}(0)$   $\tau=0$

(6)  $E[X(t)X(t+\tau)] = R_{XX}(\tau) \text{ACF} = \begin{cases} E[X(t)]E[X(t+\tau)] = \mu_X^2 & (\tau \neq 0) \\ E[X^2(t)] = R_{XX}(0) & (\tau = 0) \end{cases}$

(7)  $\text{cov}[X(t)X(t+\tau)] = R_{XX}(t, t+\tau) - \mu_X(t)\mu_X(t+\tau)$

$$C_{XX}(\tau) = R_{XX}(\tau) - \mu_X^2 = \begin{cases} 0 & \tau \neq 0 \\ R_{XX}(0) - \mu_X^2 & \tau = 0 \end{cases}$$

$$R_{X_X}(\tau) = \begin{cases} \mu_X^2 & \tau \neq 0 \\ R_{XX}(0) & \tau = 0 \end{cases}$$

### Important point :

(1) If  $X(t)$  is zero mean WSSRP.

$$E[X(t)] = 0$$

$$\sigma_{X(t)}^2 = E[X^2(t)]$$

$$\text{Var} [X(t)] = \text{MSV}\{X(t)\}$$

$$\text{Var} \{X(t=t_1)\} = \text{MSV}\{X(t=t_1)\}$$

(2) If  $X(k)$  is zero mean WSSRP

$$E[X(k)] = 0 \quad \sigma_{X(k)}^2 = E[X^2(k)]$$

$$\text{Var} [X(k)] = \text{MSV}[X(k)]$$

➤  $X(t)$ : WSSRP+IIDRP

$$E[X(t)] = \mu_X, \quad E[X(t+\tau)] = \mu_X, \quad E[X^2(t)] = R_{XX}(0) = \text{Constant}$$

$$\sigma_{X(t)}^2 = \text{Constant}$$

$$\text{Let } Y(t) = X(at+b)$$

$$E[Y(t)] = \mu_X, \quad E[Y^2(t)] = R_{XX}(0)$$

$$\sigma_{Y(t)}^2 = R_{XX}(0) - \mu_X^2$$

$$E[Y(t)Y(t+\tau)] = R_{XX}(a\tau) = R_{YY}(\tau)$$

$$\text{Cov} [Y(t)Y(t+\tau)] = C_{YY}(\tau) = R_{XX}(a\tau) - \mu_X^2$$

$$\text{For } Y(t) \rightarrow \begin{cases} \rightarrow \mu_Y = \mu_X = \text{Constant} \\ R_{YY}(\tau) = R_{XX}(a\tau) \end{cases}, Y(t) \rightarrow \text{WSSRP}$$

➤ Time shift, Time reversal, time scaling does not affect stationary nature of R.P.

$$\text{Let } Y(t) = aX(t) + b, \quad X(t) \text{ is WSSRP}$$

$$E[Y(t)] = a\mu_X + b = \text{Constant}$$

$$E[Y^2(t)] = a^2 R_{XX}(0) + b^2 + 2ab\mu_X = \text{Constant}$$

$$\sigma_{Y(t)}^2 = a^2 \sigma_{X(t)}^2$$

$$\text{Cov} [Y(t)Y(t+\tau)] = R_{YY}(\tau) - \mu_Y^2$$

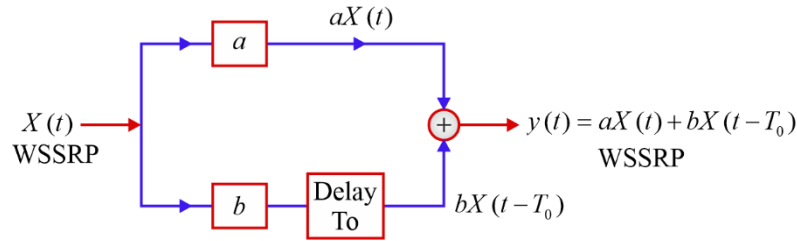
$$E[Y(t)Y(t+\tau)] = a^2 R_{XX}(\tau) + 2ab\mu_X + b^2 = R_{YY}(\tau)$$

$$y(t) \rightarrow \text{WSSRP}$$

➤ Linear transformation of WSSRP does not change its stationarity.

➤ If WSSRP passed through LTI system, output is also a WSSRP.





$$E[Y(t)] = (a + b)\mu_X$$

$$E[Y^2(t)] = a^2 R_{XX}(0) + b^2 R_{XX}(0) + 2ab R_{XX}(\tau_0)$$

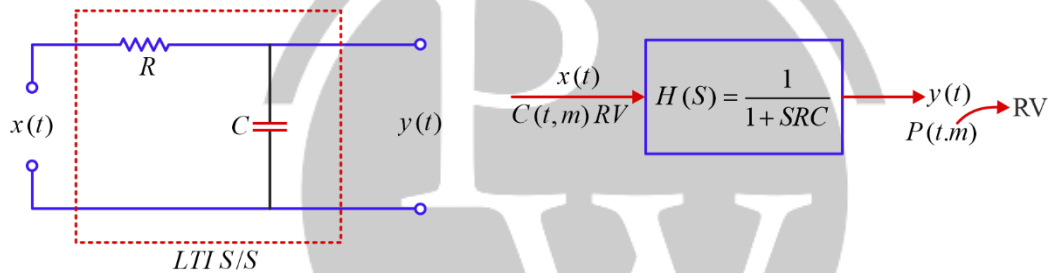
$$\sigma_{Y(t)}^2 = (a^2 + b^2)\sigma_{X(t)}^2 + 2ab[R_{XX}(T_0) - \mu_X^2]$$

$$R_{YY}(\tau) = (a^2 + b^2)R_{XX}(\tau) + abR_{XX}(\tau - T_0) + abR_{XX}(\tau + T_0)$$

$$C_{YY}(\tau) = a^2 C_{XX}(\tau) + b^2 C_{XX}(\tau) + abR_{XX}(\tau - T_0) + abR_{XX}(\tau + T_0) - 2ab\mu_X^2$$

$$E[X(n)X(n+k)] = \delta[k] = R_{XX}(k) \text{ IIDRP}$$

$$E[X(n)X(n+k)] = E[X^2(n)](k=0)$$



A,  $\omega_0 \rightarrow \text{constant}$ ,  $\theta \sim U(0, 2\pi)$  OR  $\theta \sim U[-\pi, \pi]$

$$X(t) = A \cos(\omega_0 t + \theta)$$

$$E[A \cos(\omega_0 t + \theta)] = 0$$

$$E[A \cos(\omega_0 t + \theta + \varphi)] = 0$$

$$E[X^2(t)] = \frac{A^2}{2}, \sigma_{X(t)}^2 = \frac{A^2}{2}$$

$$E[X(t)X(t+\tau)] = \frac{A^2}{2} \cos \theta \omega_0 \tau = R_{XX}(\tau)$$

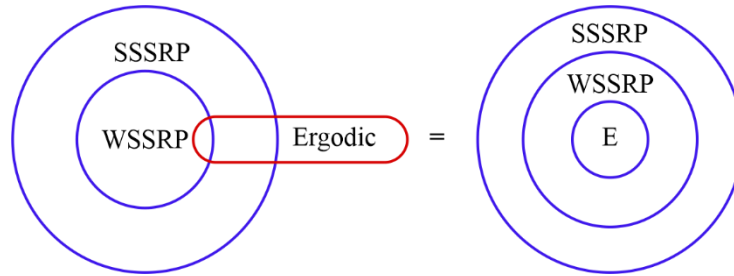
$$\text{Cov}[X(t)X(t+\tau)] = \frac{A^2}{2} \cos \omega_0 \tau$$

$X(t)$ : WSSRP + periodic with  $\tau_0 \rightarrow R_{XX}(\tau)$  will also be periodic with same T.P.

### ERGODIC Random Process :

Time Avg = Statistical Aug.

$$\frac{1}{T} \int X(t) dt = E[X(t)]$$



### Auto Correlation and its properties

Similarity between 2 Samples

Let  $X(t)$  is WSSRP,  $X(t)$  is observed  $\tau$  duration apart

$$(1) E[X(t)X(t+\tau)] = R_{XX}(\tau)$$

$$(2) R_{XX}(-\tau) = R_{XX}(\tau) : \text{Even}$$

$$(3) |R_{XX}(\tau)| \leq R_{XX}(0)$$

$$(4) \{R_{XX}(\tau)\}_{\max} = R_{XX}(0) = \text{Maximum similarity}$$

$$(5) \int_{-\infty}^{\infty} R_{XX}(\tau) d\tau = 2 \int_0^{\infty} R_{XX}(\tau) d\tau = 2 \int_{-\infty}^0 R_{XX}(\tau) d\tau$$

$$(6) \begin{cases} \text{Periodic} \rightarrow R_{xx}(\tau) : \text{Periodic} \\ \text{Non Periodic} \rightarrow R_{xx}(\tau) : \text{Non Periodic} \end{cases}$$

$$(7) E[x^2(t)] = MSV[x(t)] = R_{XX}(0) \begin{cases} \text{Energy of R.P.} \rightarrow \text{Energy Signal } x(t) \\ \text{Power of R.P.} \rightarrow \text{Power Signal } x(t) \end{cases}$$

$$(8) X(t) \text{ is power signal}$$

$$R_{XX}(0) = E[X^2(t)] = \sigma_{X(t)}^2 + \sigma_{X(t)}^2$$

$\swarrow$                        $\swarrow$                        $\swarrow$   
 Total power      A.C power      D.C power  
 of R.P              of R.P              of R.P

$$(9) \text{ If } X(t) \text{ is ergodic and WSSRP, it has no periodic component}$$

$$E[X(t)] = \mu_X \neq 0$$

$$\mu_X^2 = \lim_{\tau \rightarrow \infty} R_{XX}(\tau) = \lim_{|\tau| \rightarrow \infty} R_{XX}(\tau)$$

$$\text{If not ergodic but WSSRP then } R_{XX}(0) = E[X^2(t)], \quad R_{XX}(\infty) \neq \mu_X^2$$

**Important point:**

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

$$X(t) \leftrightarrow X(f)$$

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

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi ft} dt$$

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

$$x(t) = \int_{-\infty}^{\infty} X(f) e^{j2\pi ft} df$$

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

$$x(0) = \int_{-\infty}^{\infty} X(f) df$$

$$x(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega) d\omega$$

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

### 3.2. Parseval Theorem

$$E_{x(t)} = \int_{-\infty}^{\infty} x^2(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X(\omega)|^2 d\omega = \int_{-\infty}^{\infty} |X(f)|^2 df$$

Density Function

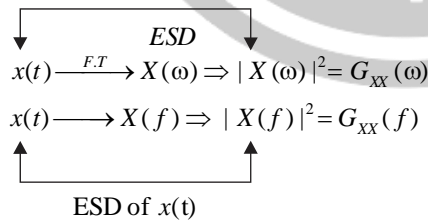
Energy spectral Density (ESD)

Power spectral Density (PSD)

$$\boxed{\begin{matrix} G_{xx}(f) \\ G_{xx}(\omega) \end{matrix}} \frac{\text{Joule}}{\text{Hz}}$$

$$\boxed{\begin{matrix} S_{xx}(f) \\ G_{xx}(\omega) \end{matrix}} \frac{\text{Watt}}{\text{Hz}}$$

Energy spectral density  $X(t) \rightarrow$  WSSRP, Engery



$$E[X(t)X(t+\tau)] = R_{XX}(\tau) \xrightarrow{F.T.} G_{XX}(\omega)$$

$$R_{XX}(\tau) \xrightarrow{F.T.} G_{XX}(f)$$

$$ACF(X(t)) \xleftarrow{F.T.} ESD[x(t)]$$

$$G_{XX}(0) = \int_{-\infty}^{\infty} R_{XX}(\tau) d\tau = 2 \int_0^{\infty} R_{XX}(\tau) d\tau$$

Zero freq. value of ESD = Area under ACF

$$R_{XX}(0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} G_{XX}(\omega) d\omega = \frac{\text{Area under ESDG}_{XX}(\omega)}{2\pi}$$

$\downarrow$   
 $E[X^2(t)]$

$$\int_{-\infty}^{\infty} G_{XX}(f) df = \text{Area under ESD} \quad G_{XX}(f)$$

### Energy Calculation :

$$E_X(t) = \int_{-\infty}^{\infty} |X(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |X(\omega)|^2 d\omega = \int_{-\infty}^{\infty} |X(f)|^2 df$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} G_{XX}(\omega) d\omega = \int_{-\infty}^{\infty} G_{XX}(f) df$$

$$G_{XX}(\omega) = G_{XX}(-\omega)$$

### Power spectral density – (PSD)

$X(t) \rightarrow$  Power signal, WSSRP

$$X(t) \xleftrightarrow{\text{PSD}} S_{XX}(\omega) = \lim_{T \rightarrow \infty} \frac{1}{T} |X_T(\omega)|^2$$

$$X(t) \xleftrightarrow{\text{PSD}} S_{XX}(f) = \lim_{T \rightarrow \infty} \frac{1}{T} |X_T(f)|^2$$

$$(1) \quad E[X(t)X(t+\tau)] = R_{XX}(\tau)$$

$$S_{XX}(\omega) = \int_{-\infty}^{\infty} R_{XX}(\tau) e^{-j\omega\tau} d\tau$$

$$(2) \quad \text{ACF}[X(t)] \xleftrightarrow{F.T.} \text{PSD}[X(t)]$$

$$R_{XX} \xleftrightarrow{F.T.} \tau_{XX}(\omega)$$

$$S_{XX}(f) = \int_{-\infty}^{\infty} R_{XX}(\tau) e^{-j2\pi f\tau} d\tau$$

$$(3) \quad S_{XX}(0) = \int_{-\infty}^{\infty} R_{XX}(\tau) d\tau = 2 \int_0^{\infty} R_{XX}(\tau) d\tau$$

Zero freq. value of = Area under ACF

PSD

$$(4) \quad R_{XX}(\tau) = \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) e^{j\omega\tau} d\omega = \int_{-\infty}^{\infty} S_{XX}(f) e^{j\omega\tau} df$$

$$R_{XX}(0) = \begin{cases} \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) d\omega \\ \int_{-\infty}^{\infty} S_{XX}(f) df \end{cases}$$

(5)  $R_{XX}(\tau) = R_{XX}(-\tau), S_{XX}(\omega) = S_{XX}(-\omega)$

(6) Calculation of power

$$E[X^2(t)] = R_{XX}(0) = \begin{cases} \frac{1}{2\pi} \int_{-\infty}^{\infty} S_{XX}(\omega) d\omega = \frac{\text{Area under PSD}}{2\pi} \\ \int_{-\infty}^{\infty} S_{XX}(f) df = \text{Area under PSD} \end{cases}$$

$$E[X^2(t)] = R_{XX}(0) = \frac{1}{\pi} \int_0^{\infty} S_{XX}(\omega) d\omega = 2 \int_0^{\infty} S_{XX}(f) df$$

### Total power

A.C. Power =  $\sigma^2 X(t)$ , D.C. Power =  $\mu^2 X(t)$

### Mean or Avg value

$$E[X(t)] = \sqrt{\frac{1}{2\pi} \int_{0^-}^{0^+} S_{XX}(\omega) d\omega}$$

$$E[X(t)] = \mu_{X(t)}^2 = \begin{cases} \frac{1}{2} \int_{0^-}^{0^+} S_{XX}(\omega) d\omega \\ \int_{0^-}^{0^+} S_{XX}(f) df \rightarrow \end{cases}$$

Is non-zero only when impulse is not present at zero frequency

$$\sigma_{X(t)}^2 = \begin{cases} \frac{1}{2\pi} \int_{-\infty}^{0^-} S_{XX}(\omega) d\omega + \frac{1}{2\pi} \int_{0^+}^{\infty} S_{XX}(\omega) d\omega \\ \int_{-\infty}^{0^-} S_{XX}(f) df + \int_{0^+}^{\infty} S_{XX}(f) df \end{cases}$$

- If  $X(t)$  is real, PSD is also real.
- PSD is even.
- PSD is non-negative,  $S_{XX}(\omega) \geq 0$ ;  $S_{XX}(f) \geq 0$

### ESD of Modulated Signal (Band Pass Signal)

$$X(t) \xleftrightarrow{\text{ESD}} G_{XX}(f)$$

↓  
Base band R.P.

$$Y(t) = X(t) \cdot A_c \cos 2\pi f_c t$$

↓  
Band pass R.P.

$$Y(f) = \frac{A_c}{2} [X(f + f_c) + X(f - f_c)]$$

Or  $\xleftrightarrow{ESD} G_{YY}(f) = |Y(f)|^2$

$$X(t) \cdot A_c \sin 2\pi f_c t \xleftrightarrow{ESD} G_{YY}(f) = |Y(f)|^2$$

$$G_{YY}(f) = \frac{A_c^2}{4} [G_{XX}(f - f_c) + G_{XX}(f + f_c)] \quad f_c \gg f_m$$

$$R_Y(\tau) = \frac{A_c^2}{2} R_X(\tau) \cos 2\pi f_c \tau$$

### PSD of Modulated Signal (Bandpass Signal)

$X(t) \rightarrow$  power single

$$Y(t) = A_c \cos(2\pi f_c t) \cdot X(t) \rightarrow S_{YY}(f) \rightarrow \text{PSD}$$

$$S_{YY}(f) = \lim_{T \rightarrow \infty} \frac{|Y_T(f)|^2}{T}$$

$$S_{YY}(f) = \frac{A_c^2}{4} \left\{ \lim_{T \rightarrow \infty} \frac{|X_T(f - f_c)|^2}{T} + \lim_{T \rightarrow \infty} \frac{|X_T(f + f_c)|^2}{T} \right\}$$

$$S_{YY}(f) = \frac{A_c^2}{4} [S_{XX}(f - f_c) + S_{XX}(f + f_c)]$$

$$R_{YY}(\tau) = \frac{A_c^2}{4} R_{XX}(\tau) \cos 2\pi f_c \tau$$

$x(t) : \text{WSSRP}$

Mixing with deterministic sinusoidal signal

Mixing with deterministic Random process

$$R_Y(\tau) = \frac{A_c^2}{2} R_X(\tau) \cos \omega_c \tau$$

$$R_Y(\tau) = \frac{A_c^2}{2} R_X(\tau) \cos \omega_c \tau$$

$$S_Y(f) = \frac{A_c^2}{2} [S_X(f - f_c) + S_X(f + f_c)]$$

$$S_Y(f) = \frac{A_c^2}{4} [S_X(f - f_c) + S_X(f + f_c)]$$

### 3.3. Cross Correlation

$X(t); \text{WSSRP}, Y(t): \text{WSSRP}$

$$E[X(t)Y(t + \tau)] = R_{XY}(t, t + \tau) = R_{XY}(\tau)$$

$$E[Y(t + \tau)X(t)] = R_{YX}(-\tau)$$

$$R_{XY}(\tau) = R_{YX}(-\tau)$$

$$\bullet \quad R_{XY}(\tau) \pm R_{YX}(-\tau)$$

- $R_{XY}(\tau) = R_{XY}(f) \rightarrow$  May/may not be
- $R_{XY}(\tau) \leq \sqrt{R_{XX}(0)R_{YY}(0)}$
- $|R_{XY}(\tau)| \leq \frac{1}{2} R_{XX}(0) + R_{YY}(0)$

### Cross Covariance

$$C_{XY}(\tau) = R_{XY}(\tau) - \mu_X \mu_Y$$

$$C_{YX}(\tau) = R_{YX}(\tau) - \mu_Y \mu_X$$

### Cross Spectral Density

$$R_{XY}(\tau) \xleftrightarrow{F.T} S_{XY}(f)$$

$$R_{YX}(\tau) \xleftrightarrow{F.T} S_{YX}(f)$$

If RP X (t) and Y (t) are orthogonal-

$$E[X(t)Y(t+\tau)] = R_{XY}(\tau) = 0 = R_{YX}(-\tau)$$

$$E[Y(t)X(t+\tau)] = R_{YX}(\tau) = 0 = R_{XY}(-\tau)$$

- X(t) and Y(t) are uncorrelated and atleast one of them have zero mean-

$$E[X(t)] = 0 \text{ or } E[Y(t)] = 0$$

$$\text{cov}[X(t)Y(t+\tau)] = 0$$

$$C_{XY}(\tau) = R_{XY}(\tau) - \mu_X \mu_Y = 0$$

$$R_{XY}(\tau) = 0$$

$$R_{XY}(\tau) = R_{YX}(\tau) = 0$$

- X(t) and Y(t) are independent R.P. and atleast one of them have zero mean.

$$\text{Cov}[X(t)Y(t+\tau)] = 0$$

$$C_{XY}(\tau) = R_{XY}(\tau) - \mu_X \mu_Y = 0$$

$$R_{XY}(\tau) = 0$$

$$R_{YX}(\tau) = 0 = R_{XY}(\tau)$$

### Combination of WSSRP

$$Z(t) = X(t) \pm Y(t)$$

$$R_{ZZ}(\tau) = R_{XX}(\tau) + R_{XY}(\tau) \pm R_{YX}(\tau) \pm R_{YY}(\tau)$$

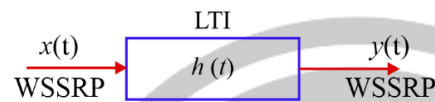
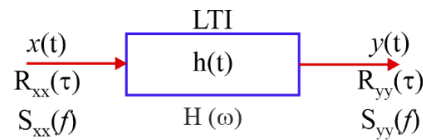
$$S_{ZZ}(f) = S_{XX}(f) + S_{YY}(f) \pm S_{XY}(f) \pm S_{YX}(f)$$

If orthogonal  $[X(t) \text{ and } Y(t)]$  then

$$R_{YX}(\tau) = R_{XY}(\tau) = 0$$

$$S_{YX}(f) = S_{XY}(f) = 0$$

### Transmission of WSSRP through in LTI system



$$E[X(t)] = \mu_X$$

$$E[Y(t)] = \mu_X [H(\omega)]_{\omega=0}$$

$$\mu_Y = \mu_X H(0)$$

If  $x(t)$  is zero mean WSSRP then  $y(t)$  is also zero mean WSSRP

$$R_{XY}(\tau) = R_{XX}(\tau) * h(\tau) \text{ and } S_{XY}(f) = S_{XX}(f)H(f)$$

$$R_{YX}(\tau) = R_{XX}(\tau) * h(-\tau) \text{ and } S_{YX}(f) = S_{XX}(f)H(-f)$$

$$R_{YY}(\tau) = R_{XX}(\tau) * h(\tau) * h(-\tau)$$

$$S_{YY}(f) = S_{XX}(f)H(f)H(-f) \quad \text{if } h(t) = \text{real}$$

$$S_{YY}(f) = S_{XX}(f) |H(f)|^2 \quad H(f) = H(-f)$$

$\downarrow$  PSD of O/P       $\downarrow$  PSD of i/p

### Power of Y(t)

$$E[Y^2(t)] = \int_{-\infty}^{\infty} S_{YY}(f) df = \int_{-\infty}^{\infty} |H(f)|^2 S_{XX}(f) df$$



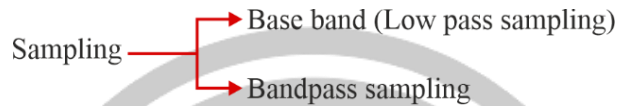


# 4

# DIGITAL COMMUNICATION

## 4.1. Sampling

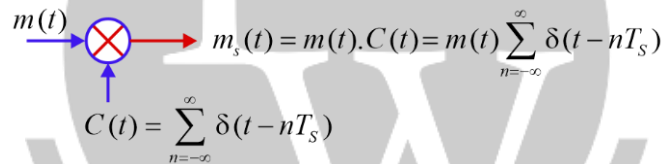
Sampling converts C.T.S into D.T.S, it retains analog or digital nature of signal.



$C(t)$ : Impulse Train – Instantaneous or Ideal sampling

$C(t)$ : Rectangular Pulse Train : Natural sampling or Flat Top sampling.

Ideal instantaneous sampling



$$m_s(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \delta(t - nT_s)$$

$$m(t) = M(f) | M(\omega)$$

$$M_s(\omega) = f_s \sum_{n=-\infty}^{\infty} M(\omega - n\omega_s)$$

$$M_s(f) = f_s \sum_{n=-\infty}^{\infty} M(f - nf_s)$$

1. If a low pass signal is sampled at  $f_s > 2f_m$  then it can be recovered from its samples, when  
 $(f_s > 2f_m) \cap (f_m \leq f_c \leq f_s - f_m)$   
 $(PBG = T_s)$   
 $f_c$  = cut off freq of ideal LPF at RX
2.  $f_s = 2f_m$ , the sampled signal  $m_s(t)$  can be recovered into  $m(t)$  if,  
 $(f_s = 2f_m) \cap (f_c = f_m)$  Ideal LPF  
 $(PBG = T_s)$

3.  $f_s < 2f_m$ , under sampling,

$T_X$  : Replace generation with ALIASING

$R_X$  : Recovery not possible

➤ ALIASING is overlapping of adjacent replica's in sampled signal.

#### 4.1.1. Low pass Sampling Theorem

A low pass sampling signal band limited to  $f_{\max}$  Hz, can be sampled and recovered from its samples when and only when

$$f_s \geq 2f_{\max} \text{ at } T_X \text{ Proper LPF at } R_X$$

No Aliasing

Recovery

#### Nyquist Rate and Nyquist Interval

Let  $m(t)$  is lowpass signal bandlimited to  $f_{\max}$  Hz.

$$f_{NY} = 2f_{\max} \quad T_{NY} = \frac{1}{f_{NY}} = \frac{1}{2f_{\max}}$$

$$(f_s)_{\min} = 2f_{\max} \quad \min \rightarrow \text{sampling rate which ensure to aliasing}$$

$$S(t) = m(t) \cos \omega_c(t) = (f_c + f_m)$$

$\uparrow$   
 $f_s(\max)$

$$N_R = 2(f_c + f_m) = f_{NY}$$

$$N_J = \frac{1}{2(f_c + f_m)} = T_{NY}$$

#### Combination of Two signals –

$$x_1(t) \rightarrow f_{\max} \rightarrow f_1, x_2(t) \rightarrow f_{\max} \rightarrow f_2$$

$$(1) \pm x_1(t) \pm x_2(t) \begin{cases} f_{\max} = \max(f_1, f_2) \\ f_{Ny} = 2f_{\max} = \max(2f_1, 2f_2) \end{cases}$$

$$(2) x_1(t) \cdot x_2(t) \begin{cases} f_{\max} = (f_1 + f_2) \\ f_{Ny} = 2f_{\max} = (2f_1 + 2f_2) \end{cases}$$

$$(3) x_1(t) * x_2(t) \begin{cases} f_{\max} = \text{Min}(f_1, f_2) \\ f_{Ny} = 2f_{\max} = \text{Min}(2f_1, 2f_2) \end{cases}$$

$$m(t) : A_m \cos \omega_m t, A_m \sin \omega_m t \rightarrow f_m$$

$$c(t) : \sum_{n=-\infty}^{\infty} \delta(t - nT_s) \rightarrow 0, f_s, 2f_s, 3f_s$$

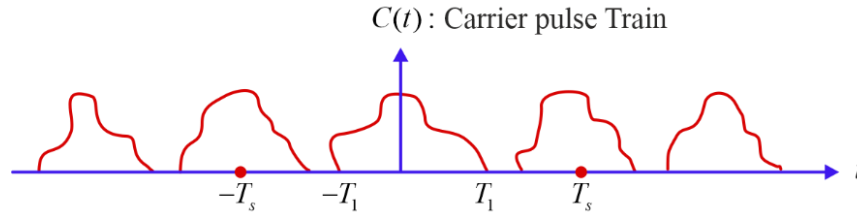
$$m_s(t) = m(t)c(t) = |0 \pm f_m|, |f_s \pm f_m|, |2f_s \pm f_m|, |3f_s \pm f_m| \dots$$

### Sampling of signal by using general carrier pulse train

$$m(t) \longrightarrow M(f) \text{ or } M(\omega)$$

$$c(t) \longrightarrow c(f) = \sum_{n=-\infty}^{\infty} C_n \delta(f - nf_s), C(\omega) = \sum_{n=-\infty}^{\infty} 2\pi C_n \delta(\omega - n\omega_s)$$

$$M_s(t) = M_s(f) = \sum_{n=-\infty}^{\infty} C_n M(f - nf_s), M_s(\omega) = \sum_{n=-\infty}^{\infty} C_n M(\omega - n\omega_s)$$

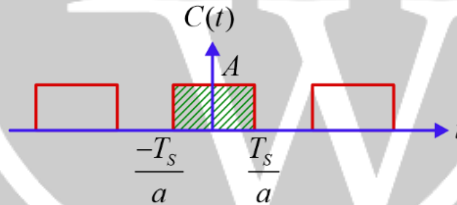


$$C_0 = \frac{\text{Area in } T_s}{T_s}$$

Recovery –  $T_X : f_s \geq 2f_m$   $R_X : \text{LPF} = \text{Proper } f_c$

$$y(t) = m(t) \quad \text{PBG} = \frac{1}{C_0}$$

If  $c(t)$  is rectangular pulse



➤  $m(t)$  is lowpass –

$$m_s(t) = c(t)m(t) \rightarrow M_s(f) = \sum_{n=-\infty}^{\infty} \frac{2A}{a} \text{sinc}\left(\frac{2n}{a}\right) M(f - nf_s)$$

Recovery-  $(f_s > 2f_m) \cap (f_m \leq f_c \leq f_s - f_m)$

PBG of LPF	$y(t)$
1	Com(t)
$\frac{1}{C_0}$	M(t)
K	$KC_0 m(t)$

$$C_0 = \frac{\text{Area in } T_s}{T_s}$$

➤  $m(t)$  is sinusoidal –

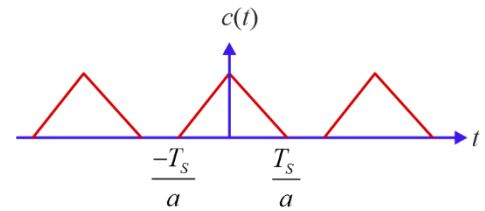
$$m(t) = A_m \cos 2\pi f_m t \longrightarrow f_m$$

$$c(t) = 0, f_s, 2f_s, \dots \text{ except } n = \frac{Ka}{2} \quad K \in I, K \neq 0$$

$$m_s(t) = 0 \pm f_m, f_s \pm f_m, 2f_s \pm f_m, \dots$$

If  $c(t)$  is Triangular-Frequency absent  $n = Ka \quad K \neq 0, K \in I$

➤ Rest of the things same as Rectangular pulse



### Bandpass Sampling :

$m(t)$  is lowpass signal.

$$f_s = \frac{2f_H}{K} \quad K = \frac{f_H}{f_H - f_L} \quad NR = 2f_H$$

### Previous Integer

$f_H$  = Maximum frequency component of Bandpass signal

### Natural Sampling

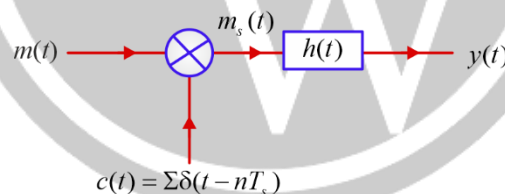
**$m(t)$ :** Low pass sampling

**$c(t)$ :** Train of finite duration pulse or rectangular pulse

$$M_s(f) = \sum_{n=-\infty}^{\infty} C_n M(f - nf_s)$$

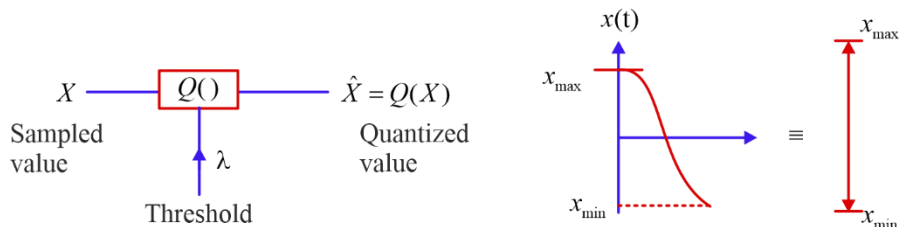
### Flat Top Sampling

Instantaneous sampling followed by a filter.



$$y(t) = \sum_{n=-\infty}^{\infty} m(nT_s) h(t - nT_s)$$

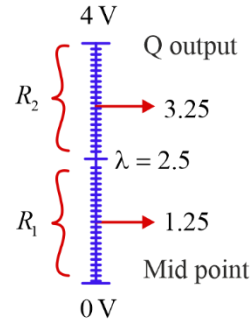
### Quantizer



Discretizes amplitude axis, analog to digital signal.

Dynamic Range of  $x(t) = (x_{\max} - x_{\min})$

$Q$  input :  $Q$  : output :  $\hat{x}(t)$



$$\hat{x}(t) = \begin{cases} 1.25 & 0 \leq x(t) \leq 2.5 \\ 3.25 & 2.5 \leq x(t) \leq 4 \end{cases}$$

➤ Many to one circuit.

### Uniform Quantizer

$$\Delta = \frac{\text{DR of } Q}{L} = \frac{m_{L+1} - m_1}{L}$$

$$\Delta_1 = \Delta_2 = \Delta_3 \dots$$

L = Number of quantization level of Q.

### Non uniform Quantizer

1.  $\Delta_1 = \Delta_2 \neq \Delta_3 \dots \Delta_i = \Delta_{i+1} \dots \neq \Delta_L$
2.  $\Delta_1 \neq \Delta_2 \neq \Delta_3 \dots \Delta_i \neq \Delta_{i+1} \dots \neq \Delta_L$

### Quantizer Error –

$$X \longrightarrow Q(\cdot) \longrightarrow \hat{X} = Q(X)$$

$$Q \cdot E = X_{QE} = X - \hat{X} = X - Q(X)$$

R.V                      Sampled Value                      Quantized Value

**Q.E.P**  $E[X_{QE}^2] = \int_{-\infty}^{\infty} x_{QE}^2 f_{QE}[qE] dqE$

When PDF of QE is given.

$$E[X_{QE}^2] = E[(X - \hat{X})^2] = \int_{-\infty}^{\infty} (x - \hat{x})^2 f_X(x) dx$$

When PDF of R.V at input of Quantizer (X) is given.

(a) PDF is uniform

$$E[X_{QE}^2] = \frac{\Delta_1^2}{12} \times A_1 + \frac{\Delta_2^2}{12} \times A_2 + \frac{\Delta_3^2}{12} \times A_3 + \dots$$

$$= \frac{\Delta_1^2}{12} [\text{Area of region in which step size is 1}] + \frac{\Delta_2^2}{12} [\text{Area of region in which step size is 2}] + \dots$$

If quantization is uniform  $E[X_{QE}^2] = \frac{\Delta^2}{12}$

(b) PDF is stair case  $E[X_{QE}^2] = \frac{\Delta_1^2}{12} \times A_1 + \frac{\Delta_2^2}{12} \times A_2 + \dots$

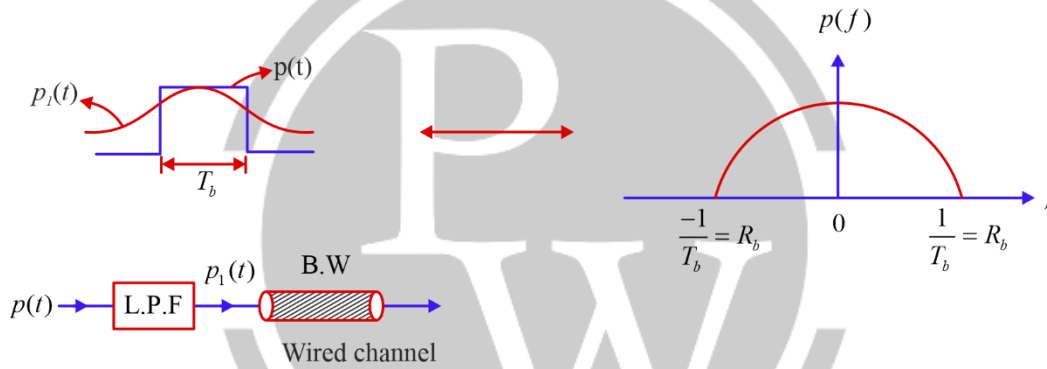
(c) PDF is non uniform  $E[X_{QE}^2] = \frac{\Delta_1^2}{12} A_1 + \frac{\Delta_2^2}{12} A_2 + \dots$

$$SQNR = \frac{\text{Signal power}}{Q \cdot E \cdot P} = \frac{E[X^2]}{E[X_{QE}^2]}$$

$$(SQNR)_{dB} = 10 \log_{10} SQNR$$

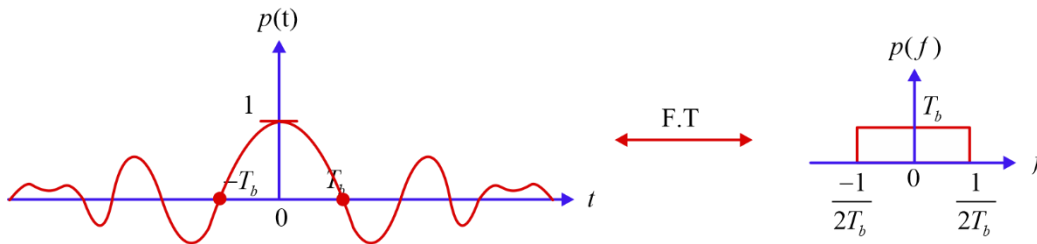
## 4.2. Pulse Transmission

### 1. Rectangular Pulse



- Bandwidth of wired channel for  $T_X$  of single pulse of duration  $T_b$  :  
 $(BW) \geq R_b$

### 2. Sinc pulse



- BW of wired channel for  $T_X$  of single  $\sin C$  pulse having zero cross over or integer multiple of  $T_b$ .  
 $(BW) \geq \frac{R_b}{2}$

$T_b$ : Bit interval

$R_b$ : Bit rate  $\rightarrow$  Bit/sec

- Minimum transmission BW of a wired channel for baseband transmission is  $= \frac{R_b}{2}$

$$\text{Number of levels } L \leq 2^n \quad L_{\max} = 2^n$$

$n$  is used to represent binary power quantization level.

### M-ary Scheme

$$M = 2^N$$

$M \Rightarrow$  Number of different symbols of duration  $NT_b$  each.

$$T_s = NT_b : \text{Symbol duration}$$

$N =$  Number of bits combined in binary sequence at a time

### Pulse Code Modulation

Bit rate,  $R_b = nf_s$  bits/sec

$$\frac{-\Delta}{2} \leq Q \cdot E \leq \frac{\Delta}{2}, \quad E(y)^2 = \frac{\Delta^2}{12}$$

If Mid point Mapping is used.

$$\Delta = \frac{\text{DR of signal}}{L} = \frac{\text{DR of } Q}{L}$$

$$L \leq 2^n \quad \frac{-\Delta}{2} \leq Q_e \leq \frac{\Delta}{2} \quad Q_e|_{\max} = \frac{\Delta}{2} \quad \Delta_{\min} = \frac{\text{DR of signal}}{2^n}$$

$$P_{QE} = E[X_{QE}^2] = E[y^2] = \int y^2 f_Y(y) dy$$

When PDF of  $Q$  is given.

$$P_{QN} = P_{QE} = \sum_{i=L}^L \int_{m_i}^{m_{i+1}} (x - \hat{x}_i)^2 f_X(x) dx$$

When PDF of  $X$  is given.

$$\text{If } Q_e \sim U\left[\frac{-\Delta}{2}, \frac{\Delta}{2}\right] = P_{QE} = \frac{\Delta^2}{12}$$

$$SQNR = \frac{12}{\Delta^2} P_s$$

$$\text{Bit Interval} \quad T_b = \frac{1}{R_b}$$

$B.W \geq R_b \rightarrow$  Rectangular Pulse

$B.W \geq \frac{R_b}{2} \rightarrow$  Since Pulse

$$(B.W)_{\min} = (BW)_{PCM} = \frac{R_b}{2}$$

### Signal to Quantization Noise Power

$m(t) \rightarrow$  Single tone sinusoidal

$$m(t) = A_m \cos \omega_m t$$

$$1. \quad P_s = \overline{m^2(t)} = \frac{A_m^2}{2}$$

$$2. \quad P_s = \overline{m^2(t)} = \frac{A_m^2}{3L^2}$$

$$3. \quad \Delta = \frac{2A_m}{L}$$

$$4. \quad SQNR = \frac{3}{2} L^2$$

$$5. \quad (SQNR)_{dB} = (1.76 + 20 \log_{10} L) \text{ dB}$$

$$6. \quad (SQNR)_{\max} = \frac{3}{2} 4^n$$

$$7. \quad (SQNR)_{\max} \text{ dB} = (1.76 + 6n) \text{ dB}$$

$x(t)$  is uniformly distributed  $[-A_m, A_m]$

$$1. \quad P_s = \frac{A_m^2}{3}$$

$$2. \quad P_{QE} = \frac{A_m^2}{3L^2}$$

$$3. \quad \Delta = \frac{2A_m}{L}$$

$$4. \quad SQNR = L^2$$

$$5. \quad (SQNR)_{dB} = 20 \log_{10} L$$

$$6. \quad (SQNR)_{dB} \leq 6n \text{ dB}$$

$$7. \quad (SQNR)_{\max} \text{ dB} = 6n \text{ dB}$$

**Key point:**

$$1. \quad SQNR = \frac{3}{2} L^2$$

$$2. \quad (SQNR)_{\max} = \frac{3}{2} 4^n$$

If  $n \rightarrow n \pm k$

$$(SQNR)_{\max} \rightarrow 4^{\pm k} \quad R_b = n f_s$$



3.  $(SQNR)_{\max} = (1.76 + 6n) \text{ dB}$

$n \rightarrow n \pm k$

$(SQNR)_{\max} \rightarrow \pm 6 \text{ dB}$

4.  $n$  given :  $L = 2^n$

5.  $L$  given:  $\begin{cases} \rightarrow L = \text{Binany Power} : L = 2^n \\ \rightarrow L \neq \text{Binany Power} : L \leq 2^n \end{cases}$

6.  $SQNR$ :  $\begin{cases} \rightarrow L = \text{Binany Power} : (SQNR) = \frac{3}{2} L^2 = (SQNR)_{\max} \\ \rightarrow L \neq \text{Binany Power} : (SQNR)_{\max} = \frac{3}{2} 4^n \end{cases}$

7.  $n$  calculation :  $n_{\min}$

8. Default  $m(t) = \text{Sinusoidal}$

### Drawback of PCM

$$BW = \frac{nf_s}{2}, P_{QE} = \frac{\Delta^2}{12}$$

$$n \uparrow \rightarrow L \uparrow \rightarrow \Delta \downarrow \rightarrow P_{QE} \downarrow \rightarrow BW \uparrow$$

$$n \downarrow \rightarrow L \downarrow \rightarrow \Delta \uparrow \rightarrow \underbrace{P_{QE} \uparrow}_{\text{}} \rightarrow BW \downarrow$$

## 4.3. DPCM (Differential Pulse Code Modulation)

### 4.3.1. PCM vs DPCM

1.  $\Delta$  fix for Both  $Q$  –

$$(BW)_{PCM} > (BW)_{DPCM}$$

$$(SQNR)_{PCM} = (SQNR)_{DPCM}$$

D.R at input of  $Q$  of PCM is greater than DPCM.

2.  $L$  fix for Both  $Q$  –

$$(BW)_{PCM} = (BW)_{DPCM}$$

$$(SQNR)_{PCM} < (SQNR)_{DPCM}$$

3. In case of DPCM the difference between current sample and predicted value of current sample is Quantized, Encoded, Line Coded and Wired  $T_{Xed}$ .

## Delta Modulator

- The recovered signal is “stair-case” approximation of  $x_{\text{ed}}$  original analog message signal.
- Stairs are added or subtracted of sampling instance.
- Size of each stair is  $\Delta$  = Step size of stairs

### Tracking Error in DM.

#### 1. Slope overload Error

$$\left| \frac{dm(t)}{dt} \right|_{\max} \gg \frac{\Delta}{T_s} \quad \text{Occurance of S.O.E}$$

- To avoid SOE,  $\Delta \uparrow \uparrow \uparrow$  by keeping  $T_s$  constant such that –

$$\left| \frac{d}{dt} m(t) \right|_{\max} \leq \frac{\Delta}{T_s} \Rightarrow \text{For sinusoidal } m(t) = A_m \cos \omega_n t$$

$$A_m \leq \frac{\Delta f_s}{\omega_m}$$

#### 2. Granular Error

It occurs when  $\Delta$  is large.

- To remove it  $\Delta \rightarrow$  small

If  $SOE \uparrow$ , G.E  $\downarrow$  and vice versa.

### SQNR in DM

$$P_{QE} = E[X_{QE}^2] = \frac{\Delta^2}{3}$$

$$1. \quad SQNR = \frac{P_s}{P_{QE}} = \frac{3P_s}{\Delta^2} \quad f_H = \text{cut off frequency of LPF}$$

$$2. \quad (SQNR)_D = \frac{3P_s}{\Delta^2} \times \frac{f_s}{f_m} = \frac{3P_s}{\Delta^2} \times \frac{f_s}{f_H}$$

$$3. \quad (SQNR)_{\max} + m(t) \text{ is sinusoidal} + \text{SOE avoid} = \frac{3}{80} \left( \frac{f_s}{f_m} \right)^2$$

$$4. \quad [(SQNR)_D]_{\max} + m(t) \text{ is sinusoidal} + \text{SOE avoid} = \frac{3}{80} \left( \frac{f_s}{f_m} \right)^3 \rightarrow \text{By default.}$$



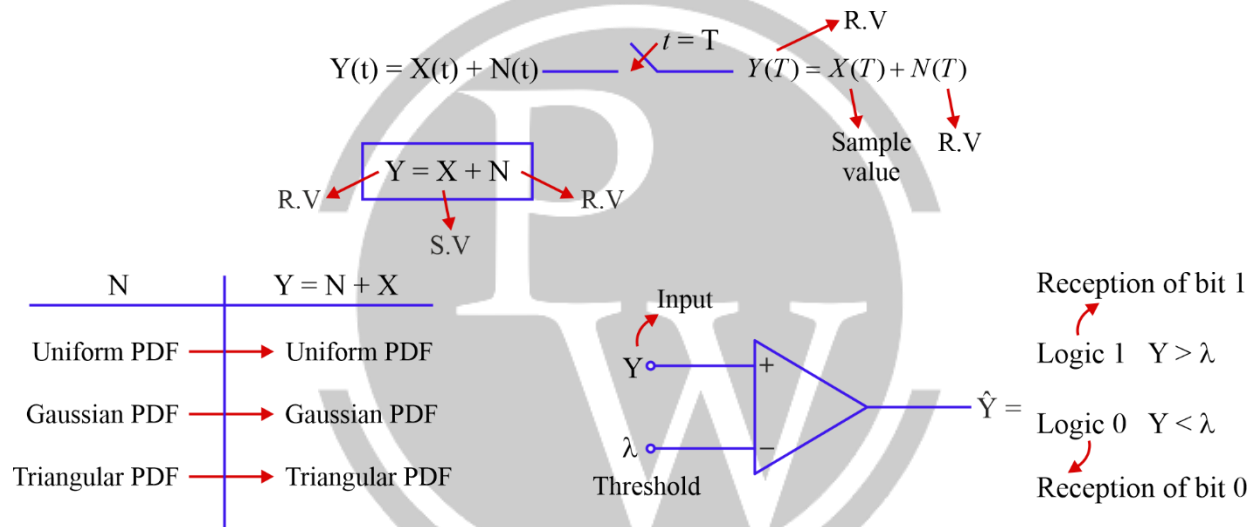
# 5

## DIGITAL RECEIVER

### 5.1. Introduction

$X(t) \rightarrow$  Deterministic signal process

$N(t) \rightarrow$  random signal process



1. If but 1 is taxed  $\rightarrow$  Receiver must recover but '1'.
2. If but 0 is taxed  $\rightarrow$  Receiver must recover but '0'.

#### Output of Sampler:

$$Y = S_{01} + N_0 = \begin{cases} S_{01} + N_{01} & 1 T_X \\ S_{02} + N_{02} & 0 T_X \end{cases} \quad \text{Channel noise is signal dependent}$$

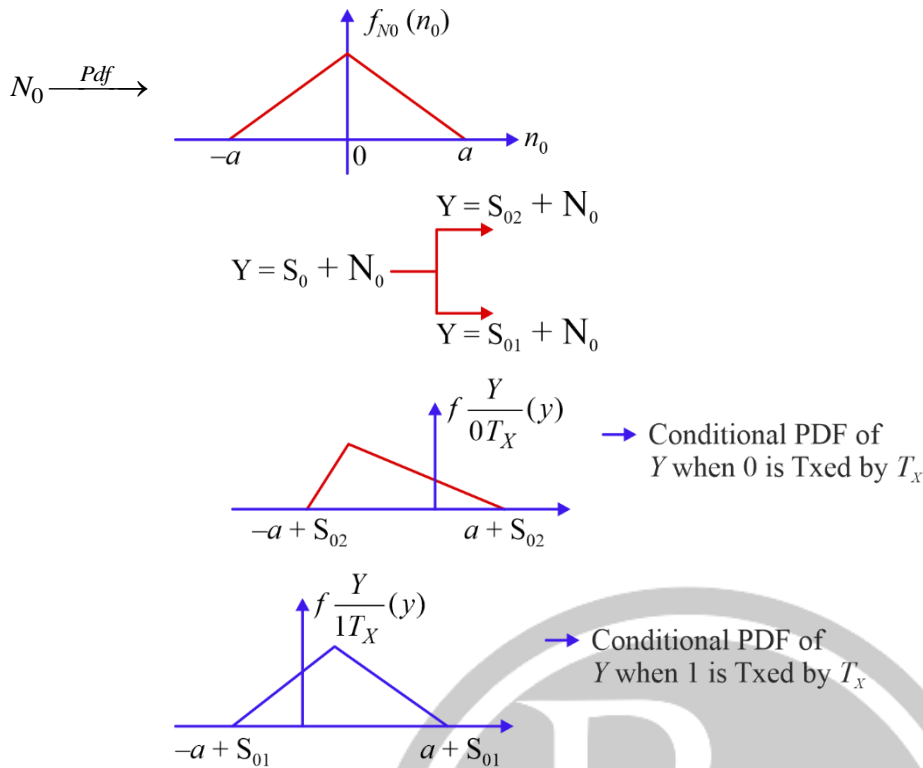
$$Y = S_0 + N_0 = \begin{cases} S_{01} + N_0 & 1 T_X \\ S_{02} + N_0 & 0 T_X \end{cases} \quad \text{Channel noise is signal independent}$$

#### BER Calculation :

$$P(1 T_X) \Rightarrow P(S_1(t): T_X) = P(S_{01}(t): \text{Reception}) = p$$

$$P(0 T_X) \Rightarrow P(S_2(t): T_X) = P(S_{02}(t): \text{Reception}) = (1 - p)$$

At the input of decision device a condition R.V. is obtained



$Y > \lambda$ : Decide in favour of 1, or decides that bit 1 would have been Txed by Txer

$Y < \lambda$ : Decide in favour of 0, or decides the bit 0 would have been Txed by Txer.

### Average Bit Error Rate :

$$P_e = P(1 T_X \cap 0) + P(0 T_X \cap 1)$$

$\downarrow$  1 Txed       $\downarrow$  decide in favour of "0"       $\downarrow$  0 Txed       $\downarrow$  decide in favour of "1"

$$P_e = P(1 T_X) P\left(\frac{0}{1 T_X}\right) + P(0 T_X) P\left(\frac{1}{0 T_X}\right)$$

$\downarrow$  decides in favour of "0" provided "1" was txed       $\downarrow$  decides in favour of "1" provided "0" was txed

### Problem Solving Technique:

**Case 1:** When PDF of Noise [noise R.V. at i/p of D.D] is given

- (1)  $\lambda$  : Given
  - (2) Calculate  $P_e = ?$
- (1) Calculate optimum threshold ( $\lambda_{opt}$ )
  - (2) Calculate min prob. of error
- Min probability of error is calculated at  $\lambda_{opt}$ .

(a)  $\lambda$  is given

$$P_e = P(0 T_X)P\left(\frac{1}{0 T_X}\right) + P(1 T_X)P\left(\frac{0}{1 T_X}\right)$$

$$P\left(\frac{1}{0 T_X}\right) = P[Y(0 T_X) > \lambda] = P(S_{01} + N_0 > \lambda) = P(N_0 > \lambda - S_{01}) = \int_{\lambda - S_{01}}^{\infty} f_{N_0}(no) dno$$

$$P\left(\frac{0}{1 T_X}\right) = P[Y(1 T_X) < \lambda] = P(S_{02} + N_0 < \lambda) = P(N_0 < \lambda - S_{02}) = \int_{-\infty}^{\lambda - S_{02}} f_{N_0}(no) dno$$

(b)  $\lambda_{opt} \rightarrow$  calculate,  $P_{e\min} \rightarrow$  calculate

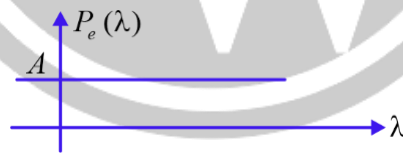
### Steps 1 :

1. Identify conditional PDF of conditional R.V. from the noise R.V. pdf (pdf of No.)

$$Y = \begin{cases} S_{01} + N_0 & 1 T_X \\ S_{02} + N_0 & 0 T_X \end{cases}$$

2. Plot the conditional PDF one over another
3. Identify the overlapping or common region and decide range of  $\lambda$   
( $\lambda_1 \leq \lambda \leq \lambda_2$ )
4. Choose any arbitrary  $\lambda$  in the above range and calculate  $P_e = P_e(\lambda)$
5.  $P_e(\lambda)$  Vs  $\lambda$

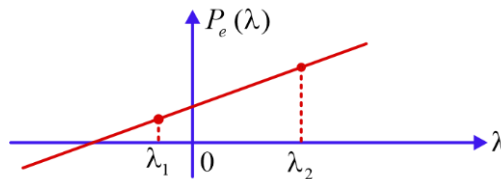
(i)  $P_e(\lambda)$  Vs  $\lambda$  : Independent of  $\lambda$



(a)  $\lambda_1 \leq \lambda \leq \lambda_2 \rightarrow$  optimum  $\lambda$  is every  $\lambda \Rightarrow \lambda \in (\lambda_1 : \lambda_2)$

(b)  $P_{e(\min)} = A$

(ii)  $P_e(\lambda)$  Vs  $\lambda$  : Linear



$\lambda_1 \leq \lambda \leq \lambda_2 \rightarrow \lambda_{opt} = \lambda_1$

$P_e(\lambda)_{\min} = P_e(\lambda_{opt})$

(iii)  $P_e(\lambda)$  Vs  $\lambda$ : Non linear

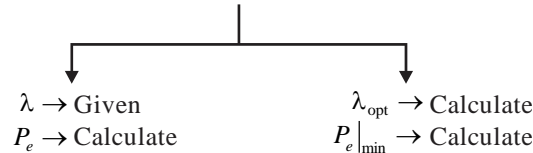
$$\boxed{\frac{d}{d\lambda} P_e(\lambda) = 0} \xrightarrow{\lambda_{opt}} \boxed{P_e(\lambda = \lambda_{opt}) = P_e|_{min}}$$

$$P_e(\lambda = \lambda_{opt}) = P_e|_{min}$$

(iv) If no overlapping region b/w conditional PDF

$$P_e(\lambda) = 0 \rightarrow \text{BER is 0}$$

**Case 2:** When PDF of conditional R.V.  $Y$  is gives at the input of the decision device

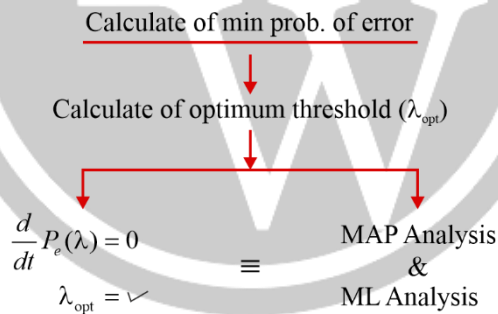


$$(a) \quad P_e = P(0 T_X)P\left(\frac{1}{0 T_X}\right) + P(1 T_X)P\left(\frac{0}{1 T_X}\right)$$

$$P\left(\frac{1}{0 T_X}\right) = P(Y(0 T_X) > \lambda) = \int_{\lambda}^{\infty} \frac{f_Y}{0 T_X}(y) dy$$

$$P\left(\frac{0}{1 T_X}\right) = P[(1 T_X) < \lambda] = \int_{-\infty}^{\lambda} \frac{f_Y}{1 T_X}(y) dy$$

(b) Same as case 1 (b)



### MAP Analysis (Maximum A posteriori Analysis)

- MAP receiver always calculate min  $P_e$ .
- Calculation of  $\lambda_{opt} \Rightarrow$  using

$$\frac{d}{d\lambda} P_e(\lambda) = 0$$

### MAP Analysis

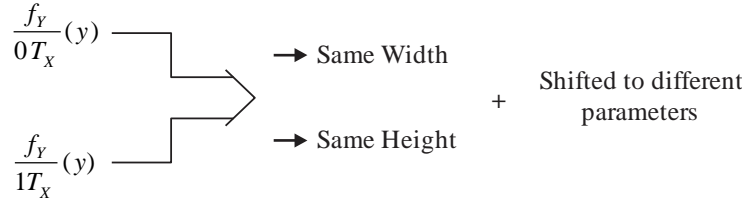
$$P\left(\frac{1 T_X}{Y}\right) \underset{"0"}{\overset{"1"}{\geq}} P\left(\frac{0 T_X}{Y}\right)$$

$$P(1 T_X) f \frac{y}{1 T_X}(y) \underset{"0"}{\overset{"1"}{\geq}} P(0 T_X) f \frac{y}{0 T_X}(y) \equiv y \underset{"0"}{\overset{"1"}{\geq}} \lambda \rightarrow \lambda_{opt} = \lambda$$

### ML Analysis : (Maximum Likelihood Analysis )

It is same as MAP analysis When  $P(0 T_X) = P(1 T_X) = \frac{1}{2}$

If noise is independent of signal then PDF of conditional R.V at input of decision device will have.



**Key point :**  $P(0 T_X) \neq P(1 T_X)$

- Noise is signal dependent/independent  $\lambda_{\min} \leq \lambda \leq \lambda_{\max}$ ,

$$\bar{\lambda} = \frac{\lambda_{\min} + \lambda_{\max}}{2}$$

$$P(0 T_X) < P(1 T_X) \quad \lambda_{opt} < \bar{\lambda}$$

$$P(1 T_X) < P(0 T_X) \quad \lambda_{opt} > \bar{\lambda}$$

- If noise is signal dependent

$$Y = \begin{cases} S_{01} + N_0 & 1 T_X \\ S_{02} + N_0 & 0 T_X \end{cases}$$

$$P(0 T_X) = P(1 T_X) = \frac{1}{2}; \lambda_{opt} = \bar{\lambda}$$

Only when  $Y$  is having Non uniform PDF.

$$P(0 T_X) \neq P(1 T_X) = \frac{1}{2}; \lambda_{opt} < \bar{\lambda}; P(0 T_X) < P(1 T_X)$$

$$\lambda_{opt} > \bar{\lambda}; P(0 T_X) > P(1 T_X)$$

### When channel noise is Gaussian Random Process

$$P_e = P(0 T_X)P\left(\frac{1}{0 T_X}\right) + P(1 T_X)P\left(\frac{0}{1 T_X}\right)$$

#### Method 1 :

$$P\left(\frac{1}{0 T_X}\right) = P[Y[0 T_X] > \lambda] = Q\left[\frac{\lambda - \mu_y[0 T_X]}{\sigma_y[0 T_X]}\right]$$

$$P\left(\frac{0}{1 T_X}\right) = 1 - Q\left[\frac{\lambda - \mu_y[1 T_X]}{\sigma_y[1 T_X]}\right]$$

Method 2 :

$$P\left(\frac{1}{0 T_X}\right) = P[y[0 T_X] > \lambda] = P[S_{02} + N_0 > \lambda] = Q\left[\frac{(\lambda - S_{02}) - \mu_{N_0}}{\sigma_{N_0}}\right]$$

$$P\left(\frac{0}{1 T_X}\right) = P[y[1 T_X] < \lambda] = P[S_{01} + N_0 < \lambda] = 1 - Q\left[\frac{(\lambda - S_{01}) - \mu_{N_0}}{\sigma_{N_0}}\right]$$

If PDF of noise at the input of D.D is given along with  $\lambda$ –

$$P\left(\frac{1}{0 T_X}\right) = P[y[0 T_X] > \lambda] = \int_{\lambda}^{\infty} \frac{f_Y}{0 T_X}(y) dy = Q\left[\frac{y - \mu_2}{\sigma}\right]$$

$$P\left(\frac{0}{1 T_X}\right) = P[y[1 T_X] < \lambda] = \int_{-\infty}^{\lambda} \frac{f_Y}{1 T_X}(y) dy = 1 - Q\left[\frac{y - \mu_2}{\sigma}\right]$$

$\lambda$  optimum

1. **Differentiation :**  $P_e = Q(\lambda), \frac{d}{d\lambda} P_e(\lambda) = 0 \rightarrow \lambda_{opt}$

2. **Map Analysis :**  $\frac{d}{d\lambda} P_e(\lambda) = 0 \rightarrow \lambda_{opt}$

$$\lambda_{opt} = \left(\frac{\mu_1 + \mu_2}{2}\right) + \frac{\sigma_y^2}{(\mu_1 - \mu_2)} \ln \frac{P(0 T_X)}{P(1 T_X)}$$

Channel noise is Gaussian, signal and channel noise are independent

$$P_e(\lambda) = P_e(\lambda_{opt}) = P_e|_{\min}$$

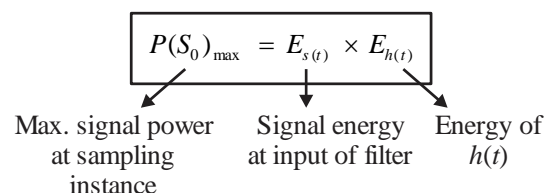
3. **ML Analysis :**  $P(0 T_X) = P(1 T_X) = \frac{1}{2}$

$$\lambda_{opt} = \frac{\mu_1 + \mu_2}{2}$$

$$P_e|_{\min} = Q\left[\frac{\mu_1 - \mu_2}{2\sigma_y}\right]$$

Schwartz Inequality

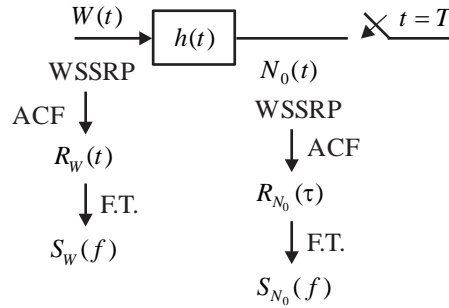
$$\left| \int_{-\infty}^{\infty} X_1(f) X_2(f) df \right|^2 \leq \int_{-\infty}^{\infty} |X_1(f)|^2 df \int_{-\infty}^{\infty} |X_2(f)|^2 df$$





$$E_{s(t)} = \int_{-\infty}^{\infty} |S(f)|^2 df$$

$$\int_{-\infty}^{\infty} |H(f)|^2 df = E_{h(t)}$$



$$P_{N_0(t)} = E[N_0^2(t)] = \frac{N_0}{2} \int_{-\infty}^{\infty} |H(f)|^2 df$$

$$E[N_0^2(t)] = \frac{N_0}{2} \times E_{h(t)}$$

$$(SNR)_{\max} = \frac{E_{s(t)}}{(N_0/2)}$$

$$\text{Only when } H(f) = [S(f)e^{j2\pi fT}]^*$$

$$(NSR)_{\max} \text{ at } t=T = \frac{\text{Energy of i/p pulse}}{\text{PSD of i/p white noise}}$$

### For General Noise

$$H(f) = [S(f)e^{j2\pi fT}]^*$$

When

$$(SNR)_{\max} = \frac{[P_{S0}]_{\max}}{P_{N_0}} = \frac{E_{s(t)} \times E_{h(t)}}{\int_{-\infty}^{\infty} S_{N_0}(f) df}$$

$$S_{N_0}(f) = |H(f)|^2 S_N(f)$$

## 5.2. Optimum Filter

$$H(f) = e^{-j2\pi fT} S^*(f) = e^{-j2\pi fT} S(-f)$$

$$h(t) = S(T-t)$$

$T$  = Sampling instance

= Duration of incoming pulse

### Unit impulse response of optimum filter

Optimum filter = matched filter  $\Rightarrow$  Maximizes signal power at sampling instances.

#### Properties of MF

$S(t)$  is an energy pulse of duration  $T$ .

1.  $h(t) = S(T - t)$
2.  $S(t) = h(T - t)$
3.  $S_0(t) = S(t) * h(t)$
4.  $S(t), h(t), S_0(t)$  are energy signal
5.  $E_S(t) = E_h(t) = |S_0(t)|_{\max}$
6.  $(SNR)_{\max}$  at  $t = T = \frac{E_s(t)}{(N_0 / 2)}$
7.  $E_{S_1}(t) > E_{S_2}(t)$  then  $Pe_1 < Pe_2$
8.  $S_0(f) = |S(f)|^2 e^{-j2\pi fT}$

$$S(t) \leftrightarrow S(f) \leftrightarrow |S(f)|^2 = G_s(f)$$

$$R_s(\tau) \xleftrightarrow{F.T.} G_s(f)$$

$$ACF[S(t)] \xleftrightarrow{F.T.} PSD[S(t)]$$

$$R_s(\tau) \xleftrightarrow{F.T.} |S(f)|^2$$

$$S_0(\tau) = R_s(\tau - T) \xleftrightarrow{F.T.} |S(f)|^2 e^{-j2\pi fT} = S_0(f)$$

#### Matched Filter Output

$$y(t) = S_0(t) + N_0(t) = \begin{cases} a_1(t) + N_0(t) & : 1 T_X \\ a_2(t) + N_0(t) & : 0 T_X \end{cases}$$

#### Channel Noise is White

$$\text{Let } P(0 T_X) = P(1 T_X) = \frac{1}{2}$$

$$[Pe]_{\min} = Q\left[\frac{a_1 - a_2}{2\sigma_y}\right] = Q\left(\frac{x}{2}\right) \quad x \uparrow \Rightarrow Q\left(\frac{x}{2}\right) \downarrow$$

Maximization of  $|x|^2$

$$|x|_{\max}^2 = \frac{\int_{-\infty}^{\infty} |S_1(f) - S_2(f)|^2 df}{(N_0 / 2)}$$

When

$$H(f) = \left[ [S_1(f) - S_2(f)] e^{j2\pi f T} \right]^*, \quad E_d = \int_{-\infty}^{\infty} |S_1(f) - S_2(f)|^2 df$$

$$x_{\max} = \sqrt{\frac{2E_d}{N_0}}$$

$$h(t) = S_1(T-t) - S_2(T-t) \rightarrow x: x_{\max} = \sqrt{\frac{2E_d}{N_0}}$$

$$P_e |_{\min} = Q\left(\frac{a_1 - a_2}{2\sigma_y}\right)$$

$$P_e |_{\min} = Q\left(\frac{x}{2}\right) \xrightarrow{h(t)} \boxed{MF} \xrightarrow{h(t)} P_e |_{\min} |_{\min} = Q\left(\frac{x_{\max}}{2}\right)$$

$$P_e |_{\min} |_{\min} = Q\left(\sqrt{\frac{E_d}{2N_0}}\right)$$

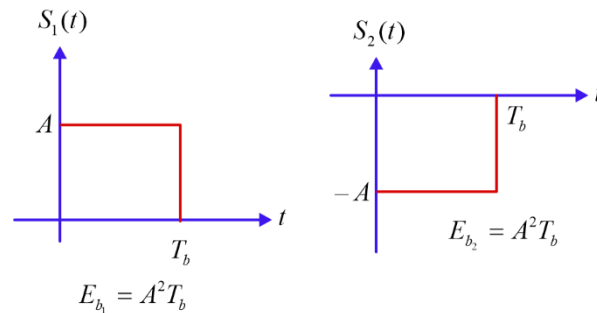
- Only when,  $P(0 T_X) = P(1 T_X) = 1/2$
- AWGN,  $\lambda \rightarrow \lambda_{opt}$
- M.F

For K noise R.V

$$P_e |_{\min} = Q\left[\sqrt{K} \left\{ \frac{(a_1 - a_2)}{2\sigma} \right\}\right]$$

$$P_e |_{\min} |_{\min} = Q\left[\sqrt{K} \sqrt{\frac{E_d}{2N_0}}\right]$$

1.



$$E_d = \int_{-\infty}^{\infty} d^2(t) dt = 4A^2 T_b$$

$$d(t) = S_1(t) - S_2(t) = 2A$$

$$0 \leq t \leq T_b$$

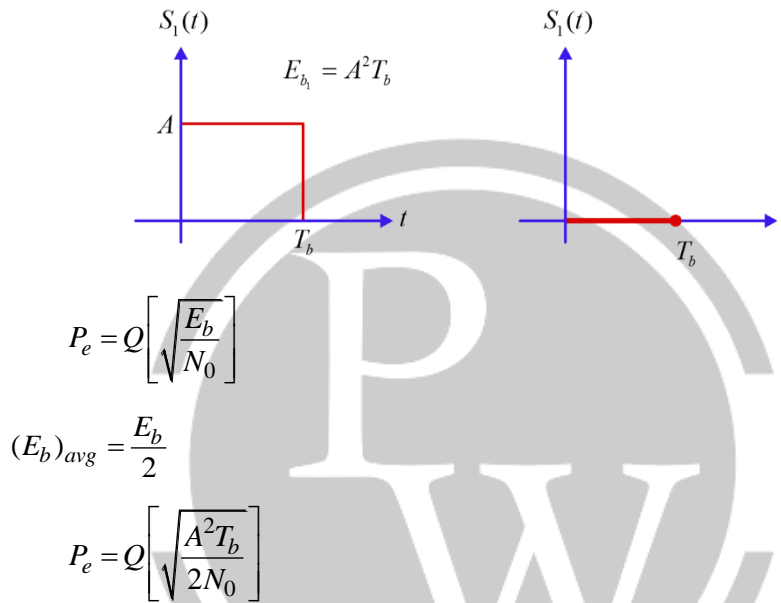
$$P_e = Q \left[ \sqrt{\frac{2A^2 T_b}{N_0}} \right]$$

$$P_e = Q \left[ \sqrt{\frac{2(E_b)_{avg}}{N_0}} \right]$$

$$(E_b)_{avg} = A^2 T_b$$

$$E_b = A^2 T_b$$

2.



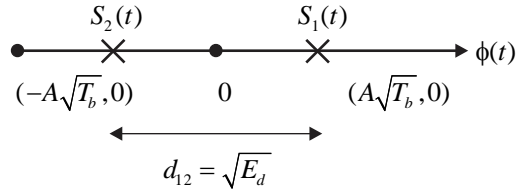
### M-ary Base Band Signaling

1. Bit rate :  $R_b$
2. Bit interval  $= T_b = 1 / R_b$
3. Symbol duration  $= T_s = NT_b$
4. Symbol rate or Baud or Baud rate  $R_s = \frac{1}{T_s} = \frac{1}{NT_b} = \frac{R_b}{N}$
5.  $T_X$  Bandwidth  $(BW) \geq R_s \rightarrow \text{Rectangular}, (BW) \geq \frac{R_s}{2} \rightarrow \sin C$

$$(BW)_{\min} = \frac{1}{2} \left( \frac{R_b}{N} \right) = \frac{R_s}{2}$$

### M-ary PAM (2-Any PAM)

1.  $M = 2, (N \leq M), s(t) = \begin{cases} S_1(t) = A & 0 \leq t \leq T_b \\ S_2(t) = -A & 0 \leq t \leq T_b \end{cases}$  NRZ coding.



$$P_e |_{\min} |_{\min} = Q \left[ \sqrt{\frac{E_d}{2N_0}} \right]$$

$$E_d = (d_{12})^2$$

$$P_e = Q \left[ \sqrt{\frac{2A^2T_b}{N_0}} \right] \text{ for NRZ}$$

$$(E_s)_{avg} = A^2T_b$$

- Distance of each point from origin =  $\sqrt{\text{Energy of that point}}$
- Distance between two point =  $\sqrt{\text{Difference energy}} = d_{12}$
- $d_{12} \uparrow \rightarrow Q(\cdot) \downarrow \rightarrow P_e \downarrow$

$$P_e = Q \left[ \sqrt{\frac{A^2T_b}{2N_0}} \right] \text{ for RZ}$$

#### Bandpass Sampling:

(a) **Binary ASK :** (m-ary ASK), For '1'  $\rightarrow A$ , '0'  $\rightarrow 0$

$$P_e = Q \left[ \sqrt{\frac{E_d}{2N_0}} \right] = Q \left[ \sqrt{\frac{A^2T_b}{4N_0}} \right]$$

$$(E_b)_{avg} = p_1E_1 + p_2E_2 = \frac{1}{2} \times \left( \frac{A^2T_b}{2} \right)$$

$$P_e = Q \left[ \sqrt{\frac{(E_b)_{avg}}{N_0}} \right]$$

$$P_e = Q \left[ \sqrt{\frac{A^2T_b \cos^2 \phi}{4N_0}} \right]$$

**Correlator Based :**

$$\phi(t) = \sqrt{\frac{2}{T_s}} \cos 2\pi f_C t$$

$$0 \leq t \leq T_s$$

$$"1" = A \cos \omega_C t \quad 0 \leq t \leq T_b$$

$$"0" = 0 \quad 0 \leq t \leq T_b$$

(b) **BPSK**:  $p(t) = \begin{cases} A & 0 \leq t \leq T_b : 1T_X \\ -A & 0 \leq t \leq T_b : 0T_X \end{cases} \rightarrow \text{Baseband}$

$$s(t) = \begin{cases} A \cos 2\pi f_c t & 0 \leq t \leq T_b : 1T_X \\ A \cos(2\pi f_c t + \pi) & 0 \leq t \leq T_b : 0T_X \end{cases}$$

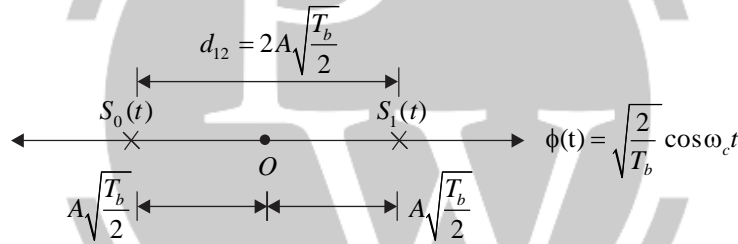
$$P_e = Q \left[ \sqrt{\frac{A^2 T_b}{N_0}} \right]$$

### Orthonormal Basis Function

$$\phi(t) = \sqrt{\frac{2}{T_b}} \cos 2\pi f_c t \quad 0 \leq t \leq T_b$$

$$P_e = Q \left[ \sqrt{\frac{A^2 T_b \cos^2 \theta}{N_0}} \right]$$

$$P_e = Q \left[ \sqrt{\frac{d_{12}^2}{2N_0}} \right]$$



### M-Ary PSK (Quadrature PSK):

$$M=4 \quad S_k(t) = A \cos \left( 2\pi f_c t + \frac{2\pi}{M} K \right) \quad 0 \leq t \leq T_s, (T_s = NT_b)$$

$$N=2 \quad S_k(t) = A \cos \left( 2\pi f_c t + \frac{\pi}{2} K \right) \quad T_s = NT_b$$

$$K=0,1,2,3 \quad d_{\min} = 2d_0 \sin \left( \frac{\phi}{2} \right) \quad d_0 = \sqrt{E_s}, \quad \phi = \frac{2\pi}{M}$$

$$d_{12} = 2d_0 \sin \left( \frac{\phi}{2} \right)$$

### M-ary PSK

$$M = (2^N)$$

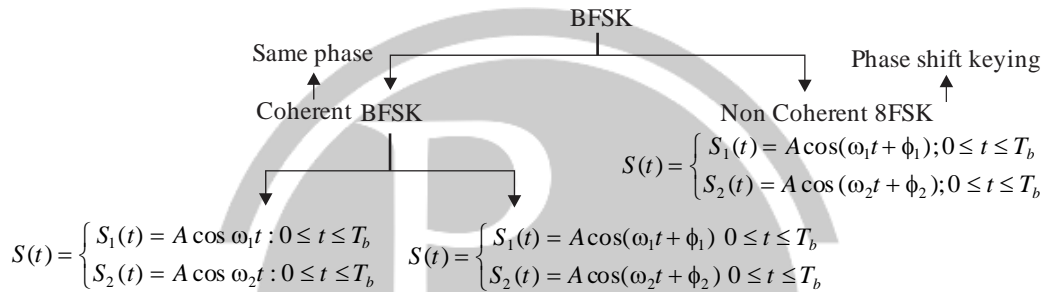
1. Bit Interval =  $T_b$ , Bit rate =  $R_b$ , symbol duration ( $T_s = NT_b$ )

2. Baud rate or symbol rate  $R_s = \frac{1}{T_s} = \frac{R_b}{N}$

3. Bit energy  $\Rightarrow E_b = \frac{A^2}{2} T_b$
4. Symbol energy  $E_s = N E_b$
5. Radius of constellation :  $d_0 = \sqrt{E_s}$
6. Area of constellation circle  $= \pi d_0^2 = \pi E_s$
7.  $d_{\min} = 2d_0 \sin\left(\frac{\phi}{2}\right), \left(\phi = \frac{2\pi}{M}\right)$

### Binary FSK

$$S_{FSK}(t) = \begin{cases} A \cos 2\pi f_1 t : 0 \leq t \leq T_b : 1T_X \\ A \cos 2\pi f_2 t : 0 \leq t \leq T_b : 0T_X \end{cases} \quad (f_1 \gg f_2)$$



### Coherent BFSK

1.  $\phi = 0, R_b = \text{HCF}[mR_b, nR_b] = \text{HCF}[2(f_1 + f_2), 2(f_1 - f_2)]$
2.  $(\phi \neq 0) R_b = \text{HCF}[mR_b, nR_b] = \text{HCF}[(f_1 + f_2), 2(f_1 - f_2)]$
3. Non-Coherent –  $R_b = \text{HCF}[mR_b, nR_b] = \text{HCF}[(f_1 + f_2), (f_1 - f_2)]$

### Condition for Orthogonality

$$\text{Coherent FSK} \begin{cases} \rightarrow d = 0(f_1 + f_2) = \frac{mR_b}{2}, (f_1 - f_2) = \frac{nR_b}{2} \\ \rightarrow d \neq 0(f_1 + f_2) = mR_b, (f_1 - f_2) = nR_b \end{cases}$$

$$R_b = \text{HCF}[mR_b, nR_b]$$

### Non-Coherent FSK

$$\phi_1, \phi_2$$

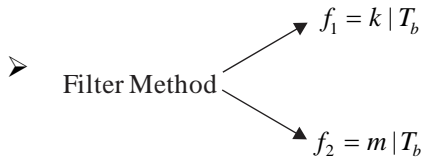
$$(f_1 + f_2) = mR_b, (f_1 - f_2) = nR_b$$

$$R_b = \text{HCF}(mR_b, nR_b)$$

$$\Rightarrow P(OT_X) = P(1T_X) = \frac{1}{2}$$

➤ Channel Noise : White (AWGN)

➤  $\lambda_{opt}$



$$Pe = Q \left[ \sqrt{\frac{E_d}{2N_0}} \right] = Q \left[ \sqrt{\frac{A^2 T_b}{2N_0}} \right]$$

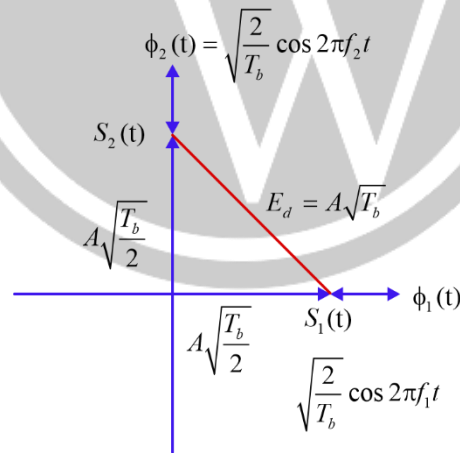
### Orthogonal FSK :

$$Pe = Q \left[ \sqrt{\frac{0.5 A^2 T_b}{N_0}} \right]$$

### Non-orthogonal FSK :

$$Pe = Q \left[ \sqrt{\frac{0.6 A^2 T_b}{N_0}} \right]$$

$$Pe = Q \left[ \sqrt{\frac{(d_{12})^2}{2N_0}} \right] = Q \left[ \sqrt{\frac{A^2 T_b}{2N_0}} \right]$$



### M-ary FSK

N bits are grouped together so that  $M = 2^N$  symbols or sinusoids of duration  $T_s = NT_b$  are generated having

➤ Same amplitude, same frequency, different frequency.

$$f_k = \frac{n}{T_s} \quad E_{s_0} = E_{s_1} = \dots = E_{s_{M-1}} = \left( \frac{A^2}{2} \times T_s \right)$$

$$T_s = NT_b \quad R_s = \frac{1}{T_s} = \frac{1}{NT_b} = \frac{R_b}{N}$$



Scheme	$P_e$	For $K$
<b>BASK</b> $\longrightarrow$	$P_e = Q \left[ \sqrt{\frac{A^2 T_b}{4N_0}} \right]$	$\Rightarrow P_e = Q \left[ \sqrt{\frac{KA^2 T_b}{4N_0}} \right]$
<b>BPSK</b> $\longrightarrow$	$P_e = Q \left[ \sqrt{\frac{A^2 T_b}{N_0}} \right]$	$\Rightarrow P_e = Q \left[ \sqrt{\frac{KA^2 T_b}{N_0}} \right]$
<b>BFSK</b> $\longrightarrow$	$P_e = Q \left[ \sqrt{\frac{A^2 T_b}{2N_0}} \right]$	$\Rightarrow P_e = Q \left[ \sqrt{\frac{KA^2 T_b}{2N_0}} \right]$
	$P_e = Q \left( \frac{\mu_1 - \mu_2}{2N_0} \right)$	$\Rightarrow Q \left[ \sqrt{K} \left( \frac{\mu_1 - \mu_2}{2N_0} \right) \right]$

For  $K$  AWGN identical independent.

### Amplitude phase shift keying (APSK)

$$S_i(t) = r_i \cos[2\pi f_c t + \theta_i] \quad (0 \leq t \leq T_s) \quad i = 0 \text{ to } \dots M-1$$

$$T_s = NT_b$$

**Case 1:**  $r_i = \text{constant}$

$\theta_i = \text{variable}$

**Case 2:**  $r_i = \text{variable}$

$\theta_i = \text{constant}$

$$\begin{cases} S_i(t) = r_i \cos(2\pi f_c t + \theta_i) \rightarrow \text{M-Ary PSK} \\ S_i(t) = r_i \cos(2\pi f_c t + \theta_i) \rightarrow \text{M-Ary ASK} \end{cases}$$

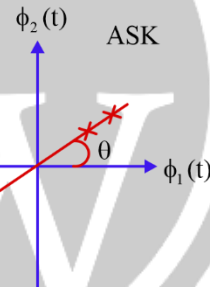


Fig. ASK

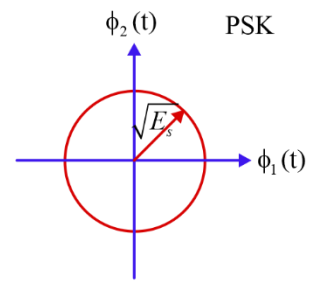
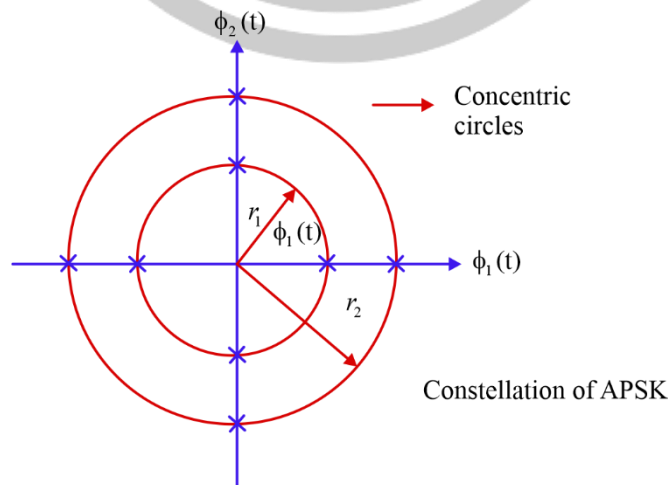


Fig. PSK



8 point APSK = 8 point QAM



# 6

# INFORMATION THEORY

## 6.1. Introduction

Information in Event ( $X = x_i$ )	Base	Unit
$I[X = x_i] = -\log_b p\{X = x_i\}$	2	Bits
	10	Decit
	$e$	Nat

### 6.1.1. Properties of Digital Information

1.  $I[X = x_i] = -\log_2 P[X = x_i]$
2.  $P[X = x_i] > P[X = x_2] \Leftrightarrow I[X = x_2] < I[X = x_i]$
3.  $P[X = 1] = \log_2 1 = 0$  bits
4.  $P[X = 0] = -\log_2 0 = \infty$  bits
5.  $0 \leq P[X = x_i] \leq 1 \Leftrightarrow 0 \text{ bits} \leq I[X = x_i] < \infty \text{ bits}$
6. For any event  $[X = x_i], I[X = x_i] \geq 0$
7.  $I[(X = x_i) \cap (X = x_2)] = I[X = x_1] + I[X = x_2]$

Average information of source  $X$  = Entropy of source  $X$

$$H[X] = -\sum_{i=1}^M P[X = x_i] \log_2 P[X = x_i] \text{ bits/symbol}$$

$$H[X] = -\sum_{i=1}^M P_i \log_2 P_i$$

**Case 1.** All  $M$  events are equiprobable –

$$H[X] = \log_2 M \text{ bits/symbol} \Rightarrow \text{Maximum entropy}$$

**Case 2.** Out of  $M$  events only 1 event is certain.

$$H[X] = 0$$

$$0 \leq H[X] \leq \log_2 M \quad M = \frac{1}{P}$$

Information Rate – Symbol rate =  $r$  symbols/sec

Entropy =  $H(X)$  bits/symbol

Information Rate  $R = r H(X)$  bits/sec

1. If  $r = f_s$  and all event equiprobable,  $L = 2^n$ ,  $H(X) = \log_2 L$

$$R = n f_s$$

### Source Coding

1. Reduces the redundancy of bits.
2. Two types of source coding
  - (a) Fixed length source coding
  - (b) Variable length source coding
    - (i) Shannon Fano coding
    - (ii) Huffman coding

### Key Point :

(a) Average code length =  $L_{avg} = \sum_{i=1}^K n_i p_i$

(b) Entropy of source =  $H(X)$

(c) Code efficiency  $\eta = \frac{H(X)}{L_{avg}}$

$\eta$  should be as high as possible.

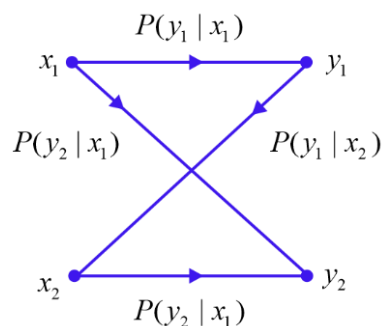
(d) Code redundancy  $\lambda = (1 - \eta)$

**Discrete channel :** A channel is called as discrete if  $X$  and  $Y$  are having finite size.

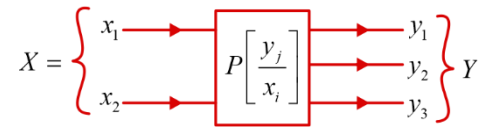
**Memoryless channel :** Each present output symbol depends on present input symbol.

$$x = \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} \rightarrow \boxed{P(y_j / x_i)} \rightarrow \begin{Bmatrix} y_1 \\ y_2 \end{Bmatrix} = y$$

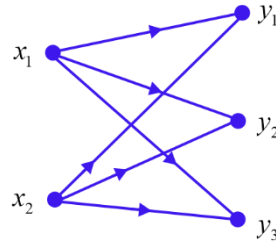
### Binary Channel : (2 input & 2 output)



### Non Binary Channel :



### Binary Channel :



\*Sum of elements of row in channel matrix is always '1'.

### Joint Channel Matrix

$$[P(x; y)] = \begin{matrix} & y_1 & \dots & y_m \\ \begin{matrix} x_1 \\ \vdots \\ x_n \end{matrix} & \begin{bmatrix} P(x_1 \cap y_1) & \dots & P(x_1 \cap y_m) \\ P(x_2 \cap y_1) & \dots & P(x_2 \cap y_m) \\ P(x_n \cap y_1) & \dots & P(x_n \cap y_m) \end{bmatrix} \end{matrix}_{n \times m}$$

$$[P(x \cap y)] = P(X)P\left(\frac{Y}{X}\right)$$

### Condition Channel Matrix

$$\left[P\left(\frac{y}{x}\right)\right] = \begin{matrix} & y_1 & \dots & y_m \\ \begin{matrix} x_1 \\ \vdots \\ x_n \end{matrix} & \begin{bmatrix} P\left(\frac{y_1}{x_1}\right) & \dots & P\left(\frac{y_m}{x_1}\right) \\ \vdots & \vdots & \vdots \\ P\left(\frac{y_1}{x_n}\right) & \dots & P\left(\frac{y_m}{x_n}\right) \end{bmatrix} \end{matrix}_{n \times m}$$

$$[P(y)] = P(X)P\left(\frac{Y}{X}\right)$$

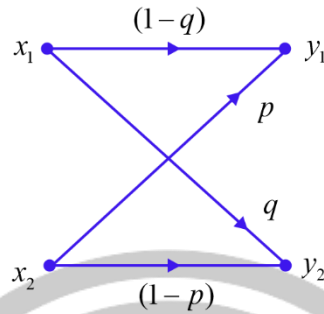
$$P(y_1) = P(x_1 \cap y_1) + P(x_2 \cap y_1) + \dots + P(x_n \cap y_1)$$

$$[P(y)]_{1 \times m} = [P(x_1), P(x_2), \dots, P(x_n)]_{1 \times n}$$

$$[P(y)]_{1 \times m} = [P(x)]_{1 \times n} \left[P\left(\frac{Y}{X}\right)\right]_{n \times m}$$

$$\begin{bmatrix} P\left(\frac{y_1}{x_1}\right) & \dots & P\left(\frac{y_m}{x_1}\right) \\ \vdots & & \vdots \\ P\left(\frac{y_1}{x_n}\right) & \dots & P\left(\frac{y_m}{x_n}\right) \end{bmatrix}_{n \times m}$$

### Binary Non-symmetrical channel



### Cross over probabilities are different

$$P\left(\frac{Y}{X}\right) = \begin{bmatrix} 1-q & q \\ p & 1-p \end{bmatrix} \begin{matrix} x_1 \\ x_2 \end{matrix}$$

$$= \begin{bmatrix} P\left(\frac{y_1}{x_1}\right) & P\left(\frac{y_2}{x_1}\right) \\ P\left(\frac{y_1}{x_2}\right) & P\left(\frac{y_2}{x_2}\right) \end{bmatrix}_{2 \times 2}$$

- $P(x_1) + P(x_2) = 1$
- $P(y_1) + P(y_2) = 1$
- $P\left(\frac{y_1}{x_1}\right) + P\left(\frac{y_2}{x_1}\right) = 1$
- $P\left(\frac{y_1}{x_2}\right) + P\left(\frac{y_2}{x_2}\right) = 1$
- $P(y_1) = P(x_1 \cap y_1) + P(x_2 \cap y_1)$
- $P(y_2) = P(x_1 \cap y_2) + P(x_2 \cap y_2)$

### Aposteriori Probabilities

$$P\left(\frac{x_1}{y_1}\right) = \frac{P(x_1)P\left(\frac{y_1}{x_1}\right)}{P(y_1)}$$

$$P\left(\frac{x_2}{y_2}\right) = \frac{P(x_2)P\left(\frac{y_2}{x_2}\right)}{P(y_2)}$$

### Map Analysis

(a) At  $r_0$ :

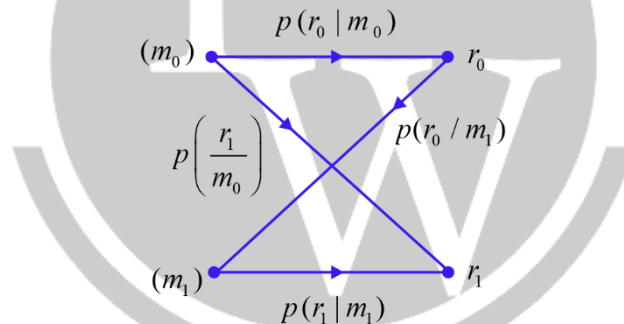
$$P\left(\frac{m_0}{r_0}\right) \underset{m_1}{\overset{m_0}{\geq}} P\left(\frac{m_1}{r_0}\right)$$

$$P(m_0)P\left(\frac{r_0}{m_0}\right) \underset{m_1}{\overset{m_0}{\geq}} P(m_1)P\left(\frac{r_0}{m_1}\right)$$

(b) At  $r_1$ :

$$P\left(\frac{m_0}{r_1}\right) \underset{m_1}{\overset{m_0}{\geq}} P\left(\frac{m_1}{r_1}\right)$$

$$P(m_0)P\left(\frac{r_1}{m_0}\right) \underset{m_1}{\overset{m_0}{\geq}} P(m_1)P\left(\frac{r_1}{m_1}\right)$$



➤ After MAP application receiver will be optimum.

### Probability of Correctness

$$P_c = P(m_0 \cap r_0) + P(m_1 \cap r_1)$$

$$P_c = P(m_0)P\left(\frac{r_0}{m_0}\right) + P(m_1)P\left(\frac{r_1}{m_1}\right)$$

$$P_e = 1 - P_c$$

### Binary Symmetrical Channel

Cross over probabilities are same.

➤  $P\left(\frac{x_1}{y_2}\right)$  = Probability that  $x_1$  was transmitted given than  $y_2$  received

$$P\left(\frac{x_1}{y_2}\right) + P\left(\frac{x_2}{y_2}\right) = 1$$

$$P\left(\frac{x_1}{y_1}\right) + P\left(\frac{x_2}{y_1}\right) = 1$$

### Joint Entropy

$$H(XY) = -\sum_{i=1}^n \sum_{j=1}^m P(x_i, y_j) \log_2 P(x_i, y_j)$$

$$H(XY) = -\sum_{i=1}^n \sum_{j=1}^m P\{(X = x_i) \cap (Y = y_j)\} \log_2 P\{(X = x_i) \cap (Y = y_j)\}$$

$$H(XY) = H(YX)$$

### Conditional Entropy

$$H\left(\frac{X}{Y}\right) = -\sum_{i=1}^n \sum_{j=1}^m P\{(X = x_i) \cap (Y = y_j)\} \log_2 P\left\{\frac{X = x_i}{Y = y_j}\right\}$$

Conditional entropy of  
X given Y

- Similarly can write  $H\left(\frac{Y}{X}\right)$

### Important point :

1.  $H(XY) = H(X) + H\left(\frac{Y}{X}\right)$
2.  $H(XY) = H(Y) + H\left(\frac{X}{Y}\right)$
3. If X and Y statically, independent  $H(XY) = H(X) + H(Y)$

$$H\left(\frac{Y}{X}\right) = H(Y), H\left(\frac{X}{Y}\right) = H(X)$$

For B.S.C –  $C_s$  = Channel capacity

$$C_s = 1 + P \log_2 P + (1 - P) \log_2 (1 - P)$$

$$C_s = \{I(X; Y)\}_{\max}$$

$$I(X; Y) = H(Y) - H\left(\frac{Y}{X}\right)$$

$$I(X; Y) = H(X) - H\left(\frac{X}{Y}\right)$$

$$H\left(\frac{Y}{X}\right) = -\sum \sum P(x_i, y_j) \log_2 P\left(\frac{y_j}{x_i}\right)$$

$$C_s = \log_2 n$$

$$I(X;Y) = H(X) \text{ loss less channel}$$

### Lossless Channel :

1. Single non zero element in each column.
2. Channel matrix should be DMC type
3. Summation of each row must be 1.
4.  $H\left(\frac{X}{Y}\right) = 0$   $I(X;Y) = H(X)$ ,  $C_s = I[X;Y]_{\max} = [H(X)]_{\max} = \log_2 n$

$n$  = number of input symbol.

## 6.2. Average Mutual Information

$$I(X;Y) = I(X) - I\left(\frac{X}{Y}\right)$$

$$I(X;Y)_{\text{Avg}} = H(X) - H\left(\frac{X}{Y}\right) \text{ bit/symbol}$$

$$I(X;Y)_{\text{Avg}} = H(Y) - H\left(\frac{Y}{X}\right)$$

$$I(X;Y) = I(Y;X)$$

$$I(X;Y) = \sum_{i=1}^n \sum_{j=1}^m P\{x_i, y_j\} \log_2 \left[ \frac{P(x_i | y_j)}{P(x_i)} \right]$$

$$I(X;Y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) \log_2 \left[ \frac{f_X\left(\frac{x}{y}\right)}{f_X(x)} \right] dx dy$$

$$P_{XY}(x_i, y_j) = P_X(y_j) P\left(\frac{x_i}{y_j}\right)$$

$$P_{XY}(x_i, y_j) = P(x_i) P\left(\frac{y_j}{x_i}\right)$$

If R.VS are Independent then  $I(x; y) = 0$



### 6.2.1. Channel Capacity

#### Maximum Average Mutual Information

$$C_s = \{I(x; y)\}_{\max}$$

$$I(x; y) = H(Y) + P \log_2 P + (1 - P) \log_2 (1 - P)$$

$$C_s = 1 + P \log_2 P + (1 - P) \log_2 (1 - P)$$

#### B.S.C

$P \rightarrow$  Cross over probability

➤ Input are equiprobable.

#### Determine Channel :

1. Number of rows in each row must be single.

- In each row angle element must be 1.
- Summation of each row will become 1.

$$H\left(\frac{Y}{X}\right) = 0, I(X; Y) = H(Y) - H\left(\frac{Y}{X}\right)$$

$$I(X; Y) = H(Y)$$

$$C_s = [H(Y)]_{\max} = \log_2 m \text{ bit/symbol}$$

#### Noise Less Channel :

1. Deterministic + Lossless

2. Each row  $\rightarrow$  Single element   
 3. Each column  $\rightarrow$  Single element   
  $\searrow$  Mult be 1

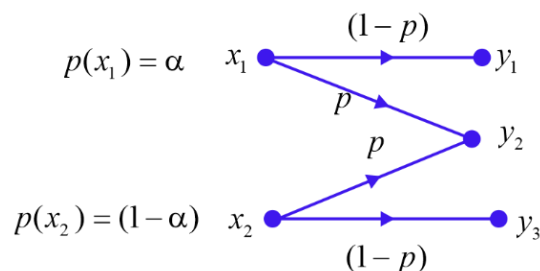
$$H\left(\frac{X}{Y}\right) = 0, H\left(\frac{Y}{X}\right) = 0$$

$$I(X; Y) = H(X) = H(Y)$$

$$C_s = [I(X; Y)]_{\max} = [H(X)]_{\max} = [H(y)]_{\max} = \log_2 m = \log_2 n \text{ bits/symbol}$$

$$m = n$$

#### Binary Erasure Channel



$$I(X;Y) = (1-P)H(X)$$

$$C_s = I(X;Y)_{\max}$$

$$= (1-P)\log_2 n \quad n=2 \text{ for BEC}$$

$$C_s = (1-P)$$

$$\Rightarrow C_s = I(X;Y) = \frac{1}{2} \log_2 \left[ 1 + \left( \frac{\sigma_x^2}{\sigma_N^2} \right) \right] \text{ Bits/symbol}$$

$$\sigma_N^2 = N_0 B$$

(i)  $X$  is zero mean R.V  $E[X^2] = \sigma_x^2 = S$

(ii) Noise is zero mean R.V  $E[N^2] = \sigma_N^2 = N$

### Channel Capacity of AWGN Channel

$$C_s = \frac{1}{2} \log_2 \left( 1 + \frac{S}{N} \right) \text{ Bit/symbol}$$

$$C = B \log_2 \left( 1 + \frac{S}{N} \right) \text{ bit/sec}$$

### Channel capacity for AWGN

$$C_s \geq \frac{R}{\text{Information rate}}$$

- For  $X$  = zero mean R.V,  $N = N_0 B$

$$B \rightarrow \infty$$

$$C_s = 1.44 \frac{S}{N_0} \text{ Finite value}$$

For loss less transmission.

## 6.3. Continuous Source and Differential Entropy

$$X : \text{DRV}, \quad H(X) = -\sum_i p[x = x_i] \log_2 p\{x = x_i\}$$

$$X : \text{CRV}, \quad H(X) = -\int_{-\infty}^{+\infty} f_X(x) \log_2 f_X(x) dx \rightarrow \text{Differential entropy}$$

$$Y : \text{DRV}, \quad H(Y) = -\sum_j p\{y = y_j\} \log_2 p\{y = y_j\}$$

$$Y : \text{CRV}, \quad H(Y) = -\int_{-\infty}^{+\infty} f_Y(y) \log_2 f_Y(y) dy \rightarrow \text{Differential entropy}$$

### 6.3.1. Channel capacity

#### (1) For error less | distortion less transmission

- (i) If all quantization level are not equiprobable :

$$C \geq R$$

$$C \geq rH(X)$$

$$C \geq f_s H(X)$$

- (ii) If all quantization level are equiprobable :

$$C \geq R$$

$$C \geq rH(X)$$

$$C \geq f_s \log_2 L$$

$$C \geq nf_s$$

$$C \geq R_b$$

#### (2) For AWGN channel- Y: GRV,

$Y = X + N$ , Let  $X$  and  $N$  are independent

$$\sigma_y^2 = \sigma_x^2 + \sigma_N^2$$

$$H(y) = \frac{1}{2} \log_2 [2\pi\sigma_y^2 e]$$

$$H(y) = \frac{1}{2} \log_2 [2\pi e (\sigma_N^2 + \sigma_X^2)] \quad \text{Maximum}$$

$$C_s = \frac{1}{2} \log_2 \left( 1 + \frac{\sigma_X^2}{\sigma_N^2} \right) \frac{\text{bits}}{\text{symbol}}$$

$P_x = \text{Power of signal } X$

$P_x = \text{Noise Power}$

$$C = C_s \times f_s$$

$$\sigma_N^2 = N_0 B$$

$$C = B \log_2 \left( 1 + \frac{\sigma_X^2}{\sigma_N^2} \right) \frac{\text{Bits}}{\text{sec}} \quad \frac{N_0}{2} \rightarrow \text{PSD of white Noise}$$

$B \rightarrow \text{B.W of channel}$

$$C = B \log_2(1 + \text{SNR})$$

↓  
Not in dB

$$\delta_{NR} = \frac{P_X}{P_N} = \frac{\sigma_X^2}{\sigma_N^2}$$

$$C = B \log_2 \left( 1 + \frac{E_b R_b}{N_0 B} \right)$$

$$C = 1.44 \frac{P_X}{N_0}$$

For infinite Bandwidth  $B \rightarrow \infty$

Information in bits | symbol

$$H \left[ \frac{y}{x_0} \right] = I \left[ \frac{y_0}{x_0} \right] P \left[ \frac{y_0}{x_0} \right] P(x_0) + I \left[ \frac{y_1}{x_0} \right] P \left[ \frac{y_1}{x_0} \right] P(x_0)$$

□□□

# 7

## MISCELLANEOUS

### 7.1. FDMA (Frequency Division Multiplexing)

- Multiple signals are multiplexed and simultaneously transmitted through channel.

$K$  = Number of signals are multiplexed

B.W  $\geq K$  [B.W of modulation scheme] +  $(K - 1)$  [BW of guard Band]

TDMA (Time division Multiplexing)

$T_s$  = Frame rate or sampling interval or time taken by commentator to complete its 1 rotation

(Band limited to same freq.)

$$T_s = nT_b \times N$$

$$T_s = NnT_b \quad N = \text{Number of signals being multiplexed}$$

$n$  = of bits/sample

$T_b$  = 1 Bit duration

$$R_b = Nnf_s$$

$$\text{Speed of commentator} = f_s \frac{\text{rotation}}{\text{sec}} = f_s \times 60 \text{rpm}$$

$$(\text{BW})_{\min} = \frac{R_b}{2} = \frac{Nnf_s}{2}$$

- When  $x$  number of synchronization  $\frac{\text{bits}}{\text{frame}}$  are added – (Band limited to same freq.)

$$T_s = (Nn + x)T_b$$

$$R_b = (Nn + x)f_s$$

- $x$  bit/frame :  $T_s = (Nn + x)T_b$

$$x \text{ bit}/2\text{frame} : T_s = \left( Nn + \frac{x}{2} \right) T_b$$

- $y\%$  (Total of  $y\%$ ) synchronization bits are added – (Band limited to same freq.)

$$T_s = \left[ Nn + \frac{Nn \times y\%}{100} \right] T_b$$

$$R_b = \left[ Nn + \frac{Nn \times y\%}{100} \right] f_s$$

- When N signals are band limited to different freq.

$$R_b = nf_{s_1} + nf_{s_2} + \dots + nf_{s_n}$$

COMA (Code division Multiple Access)

$$\text{Processing gain of CDMA} \Rightarrow G = \left( \frac{R_c}{R_h} \right)$$

- Each user is assigned with unique code

## Noise

- (1) PSD of thermal noise is Gaussian in nature. Also known as Johnson noise

- (2) Thermal Noise power  $P_n = 4KTBR = \overline{V_n^2} = (V_n)^2_{\text{rms}}$

Thermal Noise voltage

$$(V_n)_{\text{rms}} = \sqrt{4KTBR}$$

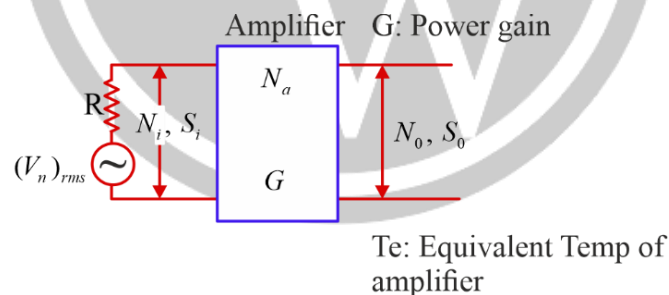
$$(I_n)_{\text{rms}} = \frac{(V_n)_{\text{rms}}}{R}$$

- Max. Power which could be deliver to amplifier = KTB
- Noise figure (F) or Noise factor

$$F(\text{dB}) = 10 \log_{10} F$$

$$N_1 = kTB$$

$$N_o = N_i G + N_a = KTBG + N_a$$



$$F = \frac{(N_i G + N_a)}{GN_i}$$

$$F = \frac{\text{Output Noise including Noisy amplifier}}{\text{Output Noise Considering noise less amplifier}}$$

$$T_e = \frac{N_a}{KBG}$$

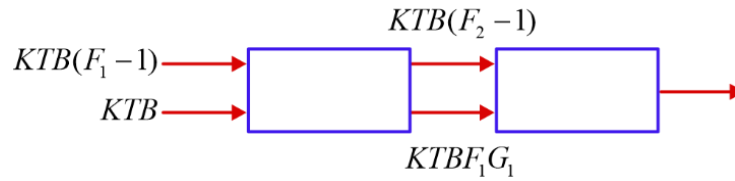
$$F = \frac{(\text{SNR})_{i/p}}{(\text{SNR})_{o/p}} = 1 + \frac{T_e}{T}$$

$$T_e = (f - 1)T$$

$$N_o (\text{output Noise power}) = KTBGF$$

$$N_o = K B G (T + T_e)$$

### Cascaded Amplifier



Output Noise with noisy  $amp^r = [KTB(F_2 - 1) + KTB F_1 G_1] G_2$

$$F = F_1 + \frac{F_2 - 1}{G_1}$$

$$F = F_1 + \frac{F_2 - 1}{G_1} + \frac{(F_3 - 1)}{G_1 G_2} + \dots$$

$$T_e \text{ (equivalent Temp.)} = T_{e1} + \frac{T_{e2}}{G_1} + \frac{T_{e3}}{G_1 G_2} + \dots$$

$$x_{dBW} = (x + 30) \text{ dBm}$$

### Noise performance of Analog Signal

$$FOM = \frac{(SNR)_0}{(SNR)_i} = \frac{SNR \text{ at the output of } R_X}{(SNR)_{in m}} = \gamma$$

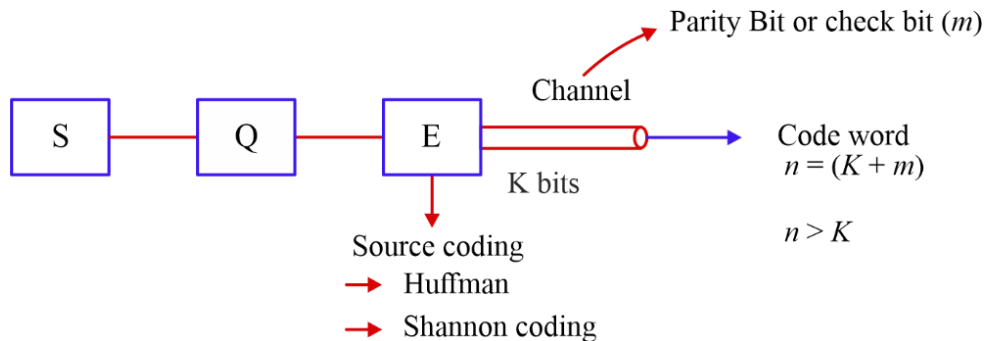
For DSB-Se  $\boxed{\gamma = 1}$ ,  $(SNR)_0 = \frac{P_m}{2N_0 B}$ ,  $(SNR)_i = \frac{P_m}{2N_0 B}$

For DCB-FC  $\boxed{\gamma = \frac{P_m}{A_c^2 + P_m}} = \eta \rightarrow \text{efficiency}$

For F.M  $\gamma = \frac{3}{4\pi^2} \frac{K_f^2 P_m}{B^2} = \frac{3}{2} \beta_{FM}^2$  (For sinusoidal)

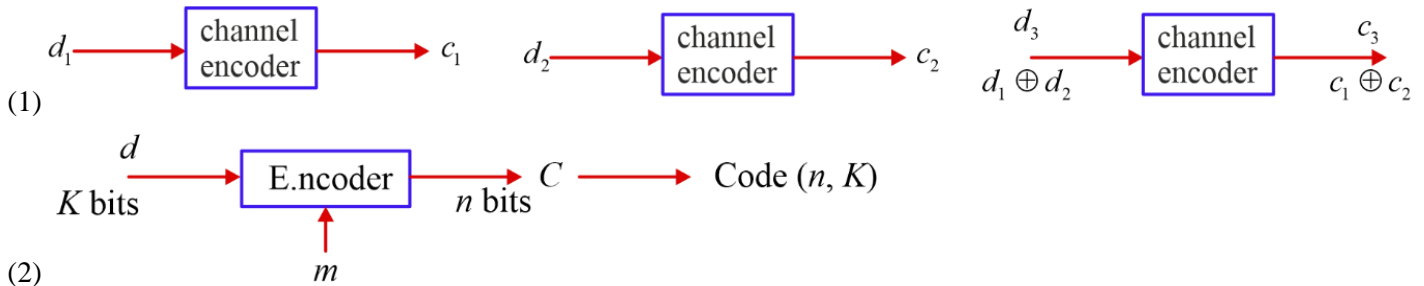
For PM  $\gamma = K_p^2 P_m = \frac{\beta_{PM}^2}{2}$  (for sinusoidal)

### Channel Coding



## Linear Block Code

$\oplus$  modulo 2 sum  $\rightarrow$  EXOR



- (3) Different data words (message word) with  $K$  bits  $= 2^K$
- (4) Each data word will have  $m$  parity bits attached to generate  $2^K$  code words.
- (5) Total no of arrangements with  $n$  bits at output of encoder will be  $\rightarrow 2^n$  out of which only  $2^K$  code words are valid.
- (6) Rate efficiency = code efficiency = code rate  $= K / n$

## Hamming Weight

Number of 1's present in L, B, C

C (7.4)

**Example:** C : 1110001, H.W = 4

## Hamming distance

It represent bit change at respective position

$$\begin{array}{ccccccc} X & = & 1 & 1 & 0 & 1 & 0 & 1 & 1 & 1 \\ & & \downarrow & \downarrow & & \downarrow & & & & \\ Y & = & 0 & 1 & 1 & 1 & 0 & 1 & 0 & 1 \end{array} \quad d(x, y) = 3$$

## Minimum Hamming Distance ( $d_{\min}$ ):

Method 1  $d_{\min} = \text{Min hamming weight of } 2^K \text{ codes except codes having 0 weight.}$

Combination:  $2^K C_2$  crosscheck

Method 2  $d_{\min} \leq n - K + 1$

Method 3 "Minimum no of columns in parity check matrix  $[H]$  Which makes zero sum (module 2)."

## Error detection L.B.C

$$d_{\min} \geq t + 1 \quad \text{can detect } t \text{ errors}$$

## Error correction $d_{\min} \geq 2t + 1$

Code Generation at  $T_X - [C]_{l \times n} = [D]_{l \times K} [G]_{K \times n} \quad \boxed{C = DG}$

$[G]_{K \times n} \rightarrow$  Generator Matrix

$[G]_{K \times n} = [I_K; p]_{K \times n}$  or  $[p; I_K]_{K \times n}$   $I_K$  = Identity Matrix of order  $K$ .

Parity check Matrix -  $[H] = [P^T; I_{n-K}]_{(n-K) \times n}$

Or

$$[H] = [I_{n-K}; P^T]_{(n-K) \times n}$$

Note -  $[C] [H^T] = 0$



### Correction at Receiver

$$c \rightarrow r$$

$$r = C \quad (\text{No error})$$

$$r \neq C \quad (\text{Error})$$

➤  $r$  will given

➤ Calculate syndrome :  $S = r[H]^T$

➤ Observe the syndrome:  $S$  matches with  $i^{\text{th}}$  row of  $[H]^T$  which Means  $i^{\text{th}}$  bit from left has error.

Non systematic L.B.C

### Hamming Code

(1) It is a L.B.C

$$(2) d_{\min} = 3$$

(3) Detect upto 2 bit error

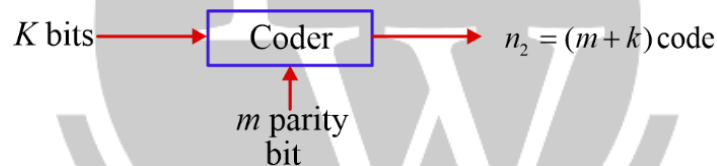
(4) Correct upto 1 bit error

(5)  $K$  bit data,  $m$  bits parity  $\Rightarrow n = (m + K)$  bits code

(6) Parity bit no. is calculated  $2^m \geq (m + K + 1)$   
 $m = ?$

(7) Placing of parity bits ate at  $2^0, 2^1, 2^2, \dots$  locations

### Cyclic redundancy check code (CRC-Code)



### Problem solving Technique:

(i)  $d = K$  bits msg

(ii) divisor polynomial

$$x^3 + x + 1 = x^3 + 0x^2 + x + 1 \rightarrow (1011)$$

**Step 1.**  $K$  msg bits are given

From  $(K + m)$  message bits

$\neq m \rightarrow$  addition of  $m$  zeros(append)

➤ Highest order of divisor polynomial (or) (number of bits in divisor polynomial)-1

**Step 2.** Modulo 2 division

□□□



**Library:-**  
**PW Mobile APP:-**

<https://smart.link/sdfez8ejd80if>  
<https://smart.link/7wwosivoicgd4>