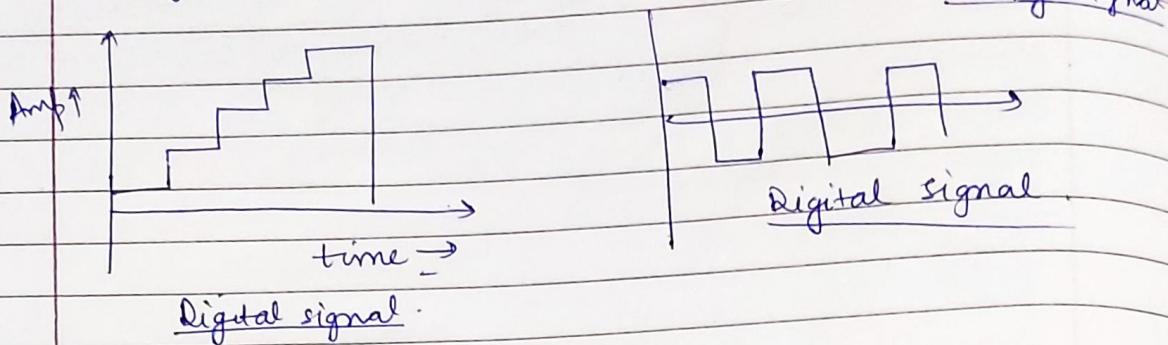


# Data Communication

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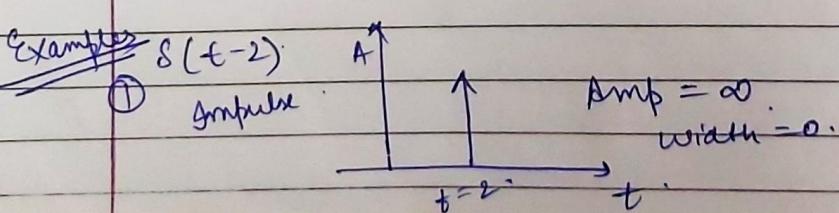
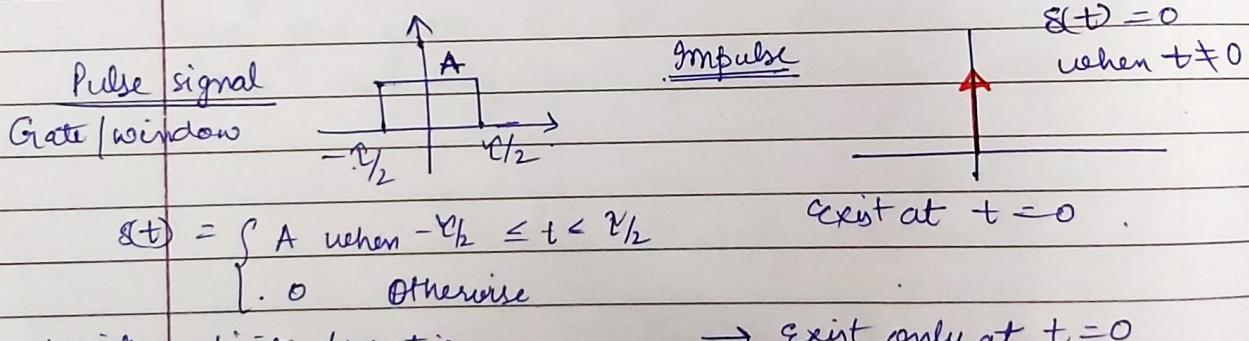
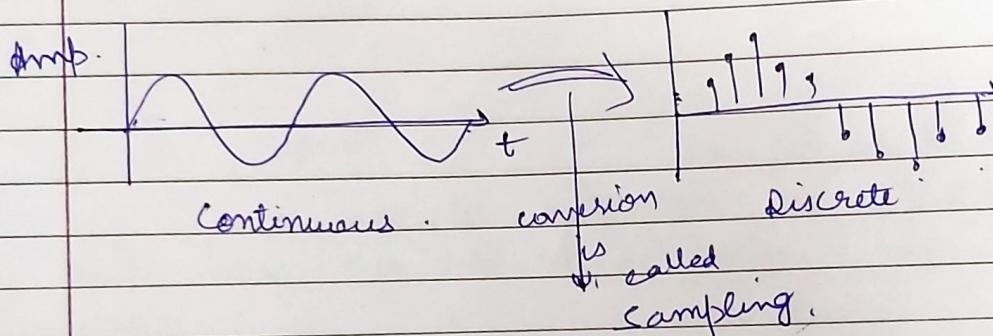
## Analog and digital Signal



~~It is discrete.~~

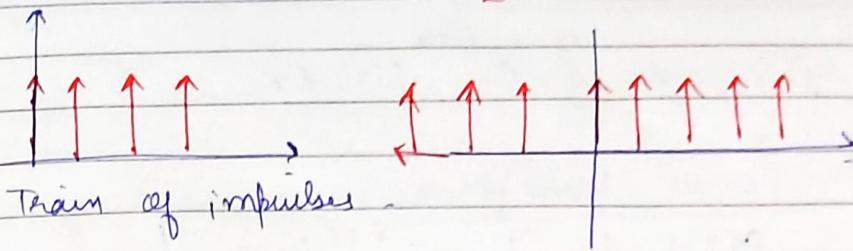
Digital Signal :- where fixed predefined amplitude ~~sa~~ labels

Discrete :- when signal exist <sup>only</sup> on specific value of time called discrete time signal.



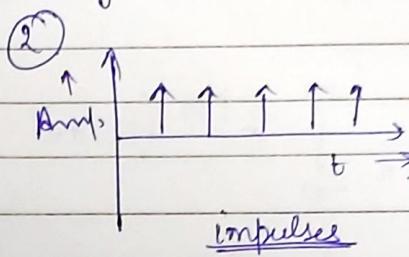
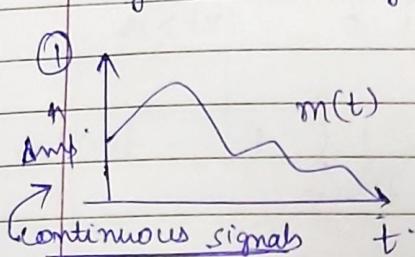
$$\textcircled{1} \quad \sum_{k=0}^{\infty} s(t-k)$$

$$\textcircled{3} \quad \sum_{k=-\infty}^{\infty} s(t-k)$$

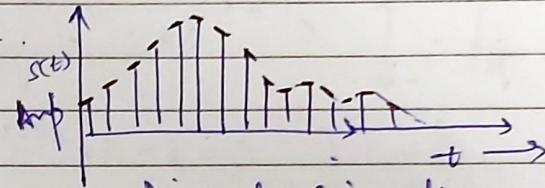


Again on sampling.

Original source signal  $m(t)$  can be of any shape and size.



\textcircled{1}  $\times$  \textcircled{2}



discrete signals ..

Fourier Transform :- Technique which converts time-domain into frequency-domain representation.

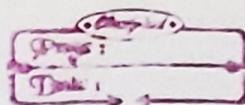
$$F(f) = \int_{-\infty}^{\infty} f(t) e^{-j 2\pi f t} dt$$

Fourier series :- is a representation of periodic signal in terms of sine and cosine harmonics

Sampling Theorem says that if the sampling frequency is greater than twice the maximum of component exist in our original signal.

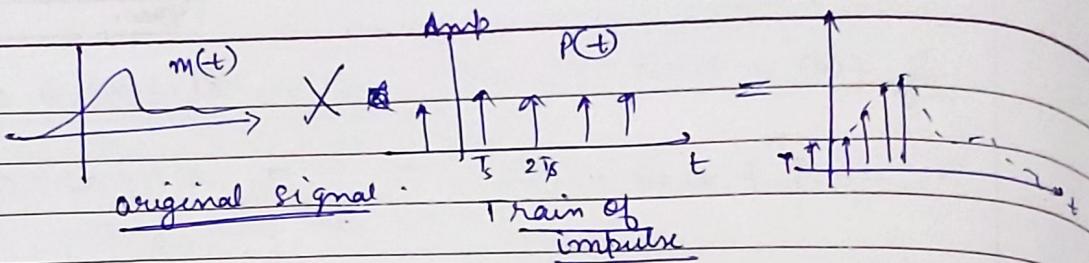
$$f_s \geq 2 f_m$$

→ Multiplicative Property of F.T.



Date

13-01-21



$$R(t) = \sum_{n=-\infty}^{\infty} s(t - nT_s)$$

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

$$\Rightarrow m(t) \xrightarrow[\text{Transform}]{\text{Fourier}} M(j\omega)$$

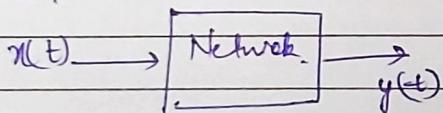
⇒  $R(t)$  : It is a periodic function. F.T. of periodic function is also train of impulse with different frequency

$$P(t) \leftrightarrow P(f) = f_s \sum_{k=-\infty}^{\infty} \delta(f - k f_s)$$

$f_s = 1/T_s$

$$\underline{H.W.} \quad C_K = \frac{1}{T} \int_{-T/2}^{T/2} g(t) e^{-j K \omega t} dt, \quad \text{if } k = \frac{1}{T_s} = f_s$$

Extra



$y(t)$  depends on  $x(t)$  and Network.

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

↑ convolution.

Acc. to Multiplicative property of F.T.

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

$$H(j\omega) = \frac{Y(j\omega)}{X(j\omega)}$$

Transfer function.

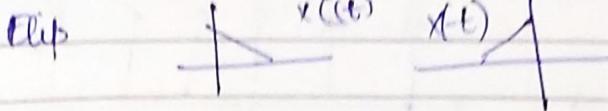
Impulse response of system

F. transform of impulse function is 1.

→ There is no existence of -ve frequency in the real world

Chap. 1  
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### Convolution



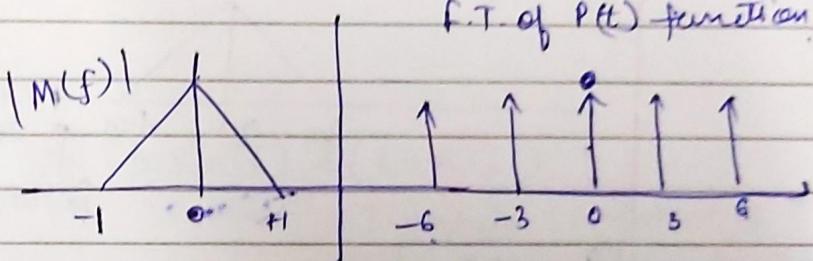
convolution }  
Flip  
Shifting  
Multiplication  
. Addition

$$x(t) * h(t) = \int_{-\infty}^{\infty} x(\tau) \cdot h(t-\tau) d\tau$$

### Back to topic

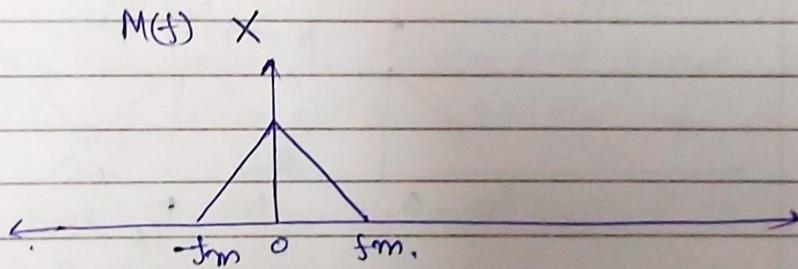
$$S(f) = f_s \sum_{k=-\infty}^{\infty} M(f - kf_s)$$

Assume  $h(t)$ .



Assumption.

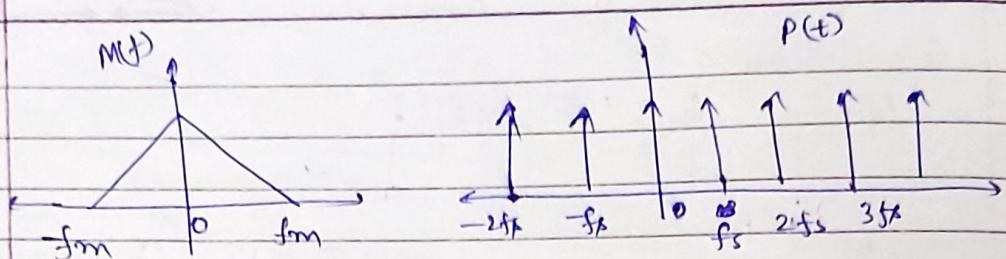
$$f_s = 2 fm$$



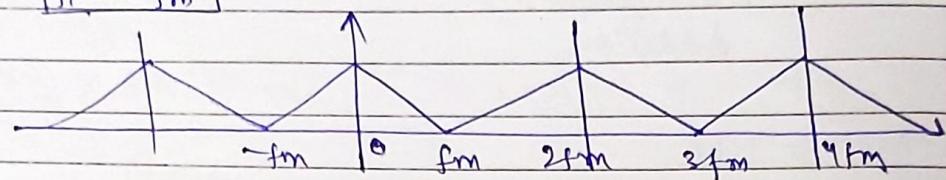
At  $k=0$

Using Multiplicative property of F.T.

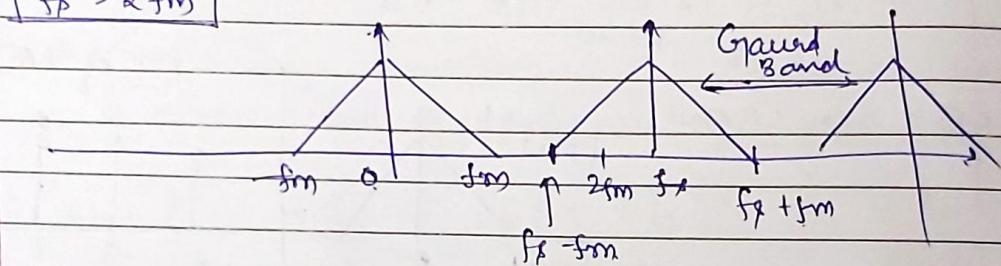
$$\left. \begin{array}{l} \text{After} \\ \text{F.T.} \end{array} \right\} \begin{array}{l} s(t) \longleftrightarrow S(f) \\ m t \longleftrightarrow M(j\omega) \\ p(t) \longleftrightarrow P(f) \quad (\text{This will give a train of impulses} \\ \text{but with different frequencies}) \end{array}$$



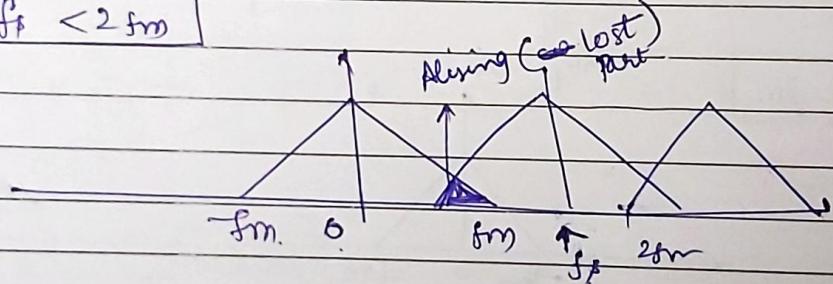
Case 1  $f_p = 2\text{ fm}$



Case 2  $f_p > 2\text{ fm}$



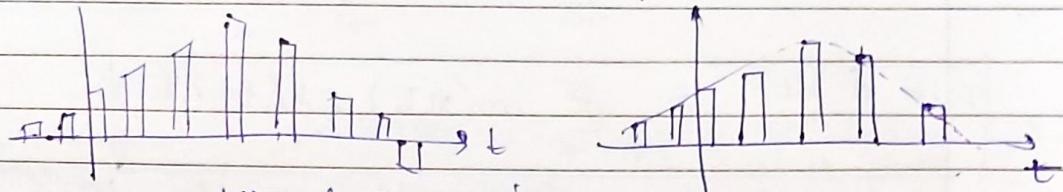
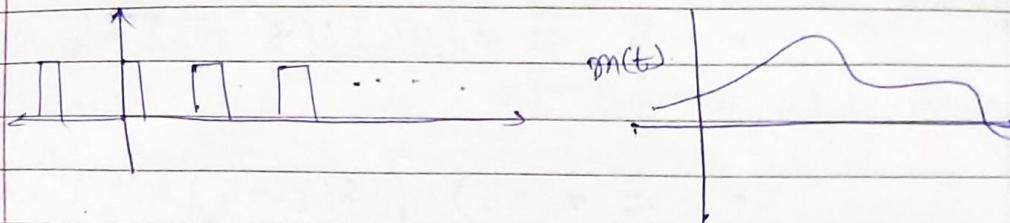
Case 3  $f_p < 2\text{ fm}$



$f_r = 2\text{ fm}$  is an ideal case (Not feasible in real world)  
So,  $f_s > 2\text{ fm}$ .

Case 4

Pulse Sampling :- we train of pulses.



Natural Sampling

flat top sampling

### Flat-Top Sampling

so,

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

↑ impulse response of  
pulse broadening function

→ We take the instantaneous sample and passed it through pulse broadening circuit to get flat top sample.  
or.

Took the instantaneous sample and kept it for more period of time

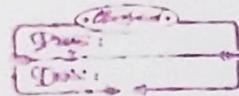
$$h(t) = \begin{cases} 1 & 0 < t < T \\ 0 & \text{elsewhere} \end{cases}$$

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

discrete time  
signal of  $m(t)$

$$s(t) \cdot x(t) = x(0)$$

$$s(t-2) \cdot x(t) = x(2)$$



$$m(t) * h(t) = \int_{-\infty}^{\infty} m(z) h(t-z) dz.$$

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

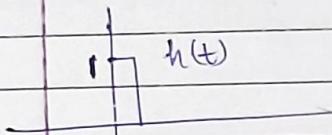
$$= \sum_{n=-\infty}^{\infty} m(nT_s) \int_{-\infty}^{\infty} \delta(t-nT_s) h(t-nT_s) dz$$

$$[m(t) * h(t) = \sum_{n=-\infty}^{\infty} m(nT_s) \cdot h(t-nT_s)]$$

for  $n=0$ :

$$m(0) \cdot h(t)$$

function  
for flat  
top sampling



### H.W. Natural Sampling

Quantization :- Process in which we restrict our function with fixed magnitude ~~levels~~ levels.

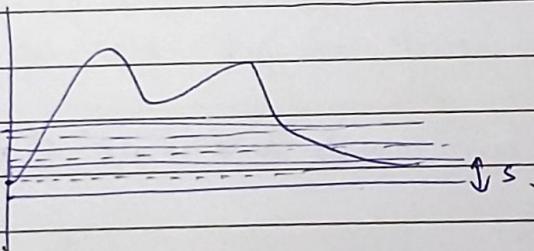
$$V_H = \text{Max. Voltage}$$

$$V_L = \text{Min. "}$$

$$S = V_H - V_L$$

$M \rightarrow$  No. of levels

$S \rightarrow$  step size



→ step size is very small

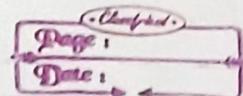
→ signal is not variable in each step -

→ value of each step is equal to middle val.

→ As quantised signal is not original signal so it will generate some error known as Quantization error.

$$\text{Quantization error} = m(t) - m_q(t)$$

Mean. Q. error =  $\frac{s^2}{12}$



- To reduce quantization error, step size should be as small as possible
- Channel noise says step size should be as large as possible
- Trade off

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15/01/21

Quantization

linear

Non-linear

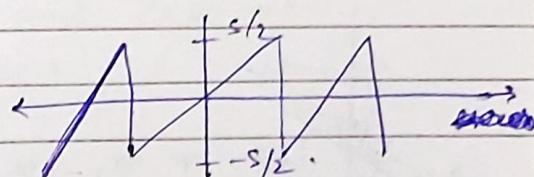
# Quantization error varies from  $s/2$  to  $-s/2$ .

### Linear Quantization

- Midtread  
→ Midrise

Quantization noise :- Diff.

b/w original signal and quantized signal.



$$f_x(x) = \begin{cases} 1/s & -s/2 \leq x \leq s/2 \\ 0 & \text{otherwise} \end{cases}$$

• Random variable  $x$ , which is uniformly distributed

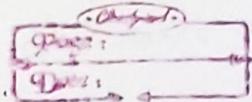
### Mean square error (Noise variance)

$$\sigma_x^2 = \int_{-\infty}^{\infty} x^2 f_x(x) dx$$

$$= \frac{1}{s} \int_{-s/2}^{s/2} x^2 dx.$$

$$= \frac{1}{s} \left[ \frac{x^3}{3} \right]_{-s/2}^{s/2} = \frac{1}{s} \times \frac{1}{3} \left( \frac{s^3}{4} \right)$$

$$\boxed{\sigma_x^2 = \frac{s^2}{12}}$$



? Quantization error : signal to noise ratio

→ 2<sup>nd</sup> way of calculating error.

let  $f(m) dm$  be the probability that  $m(t)$  lies in a range  $m - dm/2$  to  $m + dm/2$

Then the mean square quantization error.

$$\overline{e^2} = \int_{m_1 - s/2}^{m_1 + s/2} f(m) (m - m_1)^2 dm + \int_{m_2 - s/2}^{m_2 + s/2} f(m) (m - m_2)^2 dm + \dots$$

$m_1$ : mid point  
of 1st step.

~~Assumptions~~  
let  $m - m_k = x$ ,

Step size small so, probability distribution is constant

$$\overline{e^2} = \left\{ f^{(1)} + f^{(2)} + f^{(3)} + \dots \right\} \int_{-s/2}^{s/2} x^2 dx .$$

$$\overline{e^2} = \left\{ f^{(1)} + f^{(2)} + \dots \right\} s^3/12 .$$

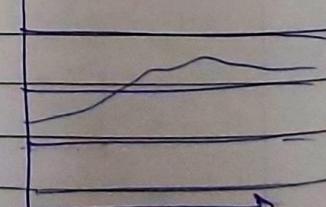
$$\overline{e^2} = \underbrace{\{ s \cdot f^{(1)} + s \cdot f^{(2)} + s \cdot f^{(3)} + \dots \}}_{\text{Probability distribution.}} \cdot s^2/12 .$$

Probability that signal  
lies in 1<sup>st</sup> step.

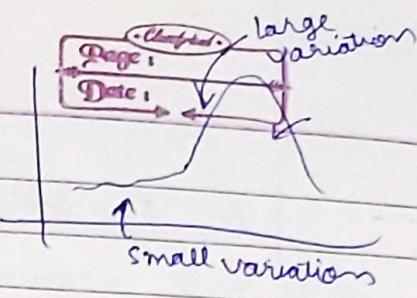
$$\overline{e^2} = 1 \cdot s^2/12 = s^2/12 .$$

# If step size is large such that  
whole function lies b/w two  
or few step. Then we will  
not get the replica of exact  
signal

Quantization noise becomes very large.



# If step is small, channel noise become very large

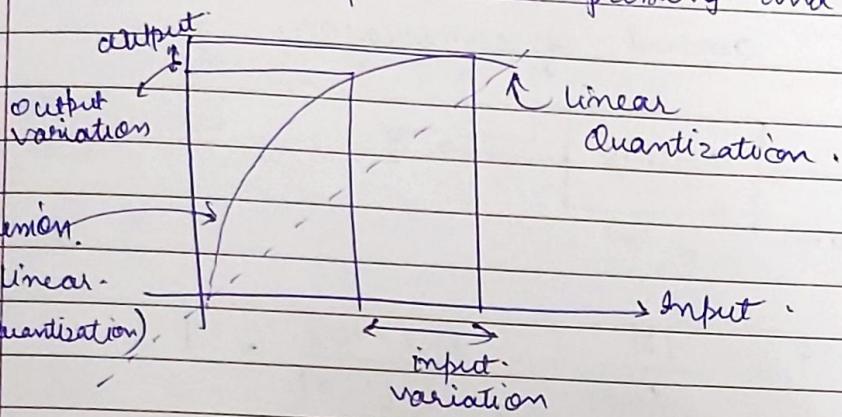


- So, for large variation we should take large step size.
- and for small variations we should take small step size.
- So, we can't take fixed step size.
- So, in ~~step~~ Non-linear Quantization. there is variable step size exist.

### Non-linear Quantization.

Companding :- compression and expansion.

- Step size keeps on compressing and expanding.



Compressor  $\rightarrow$  Uniform Quantizer  $\rightarrow$  Expander.  
For getting  $\rightarrow$  Compressor

Non-linear Quantization  $\rightarrow$  Uniform Quantizer  
 $\rightarrow$  Expander.

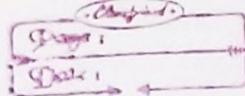
Laws (In non-linear Quantization)

① A-Law

② μ-Law

Application of Sampling :- Can send more than one signals simultaneously

from youtube



## μ - Law Companding

compression + expansion  
(at input side)      (at

(at input)

(at output side)

$\rightarrow$   $y_p \rightarrow o/p$  relationship is given by

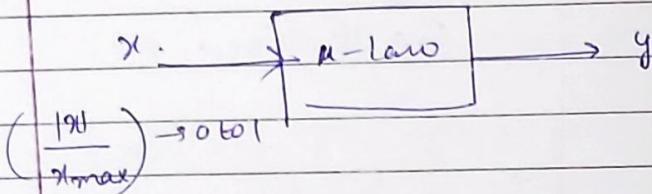
$$\frac{|y|}{x_{\max}} = \frac{\ln \left[ 1 + \mu \frac{1-x}{x_{\max}} \right]}{\ln [1+r]}$$

where,

$x$  = Amp. of 1/p signal at a particular instant

$y$  = compressed o/p.

$\mu$  = unitless parameter used to define the amount of compression.



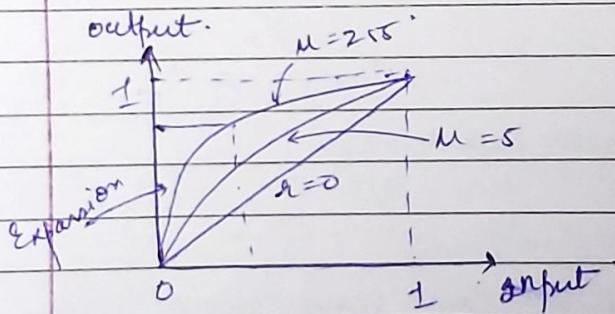
$$\text{for } u=0, \frac{|y|}{x_{\max}} = \frac{\ln[1+0]}{\ln[1+0]} = 1$$

$$|y| = x_{\max}.$$

↳ un-compressed output

→ There is no comprehension.

as  $n \uparrow$  compression will happen.



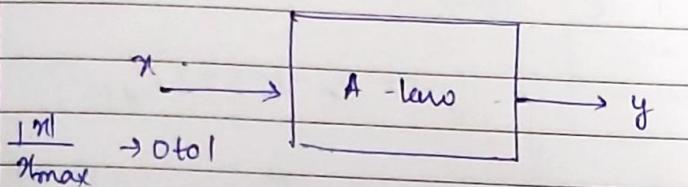
→ for smaller values of input, output variation is large and for larger values of input, output variation is small.

→ Larger value of  $n$ , more is the compression

## A-law companding

- It has slightly flatter o/p characteristic than u-law
- I/O → O/P relationship is given by.

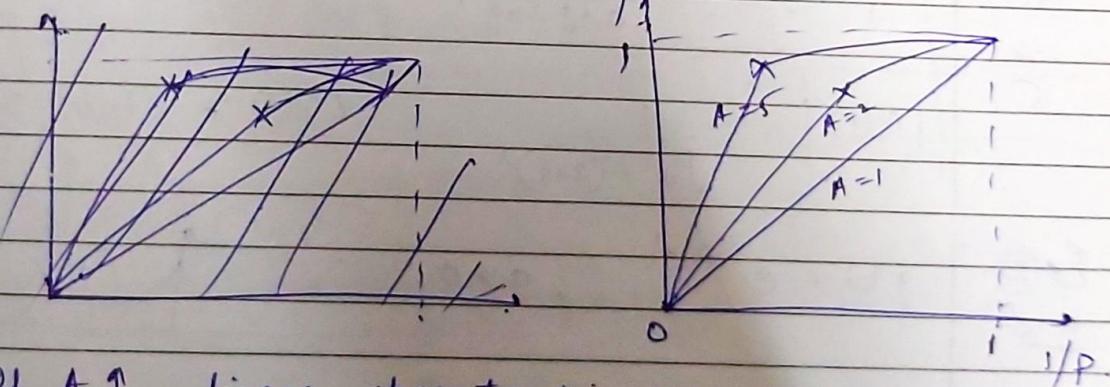
$$\frac{|y|}{x_{\max}} = \begin{cases} \frac{A|x|}{x_{\max}} & ; 0 \leq \frac{|x|}{x_{\max}} \leq \frac{1}{A} \\ \frac{1 + \ln \left[ \frac{A|x|}{x_{\max}} \right]}{1 + \ln A} & ; \frac{1}{A} \leq \frac{|x|}{x_{\max}} \leq 1 \end{cases}$$



$$\rightarrow \text{If } A = 1 \quad \frac{|y|}{x_{\max}} = \frac{A|x|/x_{\max}}{1 + \ln A} = \frac{|x|}{x_{\max}}$$

$$|y| = |x|$$

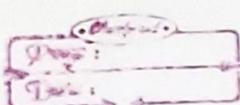
→ As  $A \uparrow$  compression  $\uparrow$  o/p



- If  $A \uparrow$ , linear characteristic shift towards left.
- for small  $x$  value  $\rightarrow$  expansion
- for larger values  $\rightarrow$  compression.

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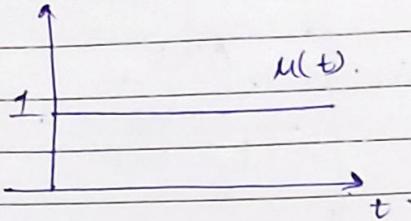


## Fourier Transform

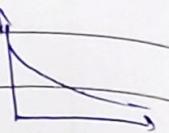
$$F\{g(t)\} = \int_{-\infty}^{\infty} g(t) e^{-j2\pi ft} dt.$$

Unit Step function.

$$u(t) = \begin{cases} 1 & \text{when } t \geq 0 \\ 0 & t < 0 \end{cases}$$



Ex (1)  $g(t) = e^{-at} u(t)$  Fourier Transform?



$$F\{g(t)\} = \int_{-\infty}^{\infty} e^{-at} u(t) e^{-j2\pi ft} dt.$$

$$= 0 + \int_0^{\infty} e^{-at} e^{-j2\pi ft} dt.$$

$$= \int_0^{\infty} e^{-(a+j2\pi f)t} dt = \left[ \frac{e^{-(a+j2\pi f)t}}{-(a+j2\pi f)} \right]_0^{\infty}$$

$$= \frac{-1}{a+j2\pi f} [0-1]$$

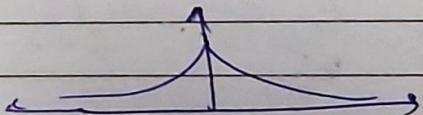
$$G(f) = F\{g(t)\} = \frac{1}{a+j2\pi f}$$

Phase angle

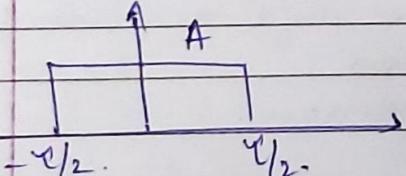
$$|G(f)| = \frac{1}{\sqrt{a^2 + (2\pi f)^2}} \quad \angle G(f) = -\tan^{-1}\left(\frac{2\pi f}{a}\right)$$

Ex (2)  $g(t) = e^{-at}, a > 0$

H.W.

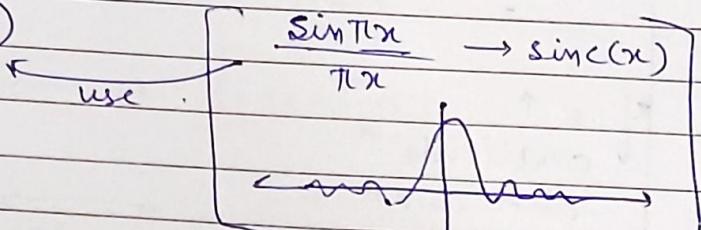


Ex (3)



$$g(t) = \begin{cases} A, & -T/2 \leq t \leq T/2 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned}
 F\{g(t)\} &= \int_{-\infty}^{\infty} g(t) e^{-j2\pi f t} dt \\
 &= \int_{-\infty}^{\infty} A \cdot e^{-j2\pi f t} dt = A \cdot \left[ \frac{e^{-j2\pi f t}}{-j2\pi f} \right] \Big|_{-\infty}^{\infty} \\
 &= \frac{A}{j2\pi f} \left[ e^{j\pi f \infty} - e^{-j\pi f \infty} \right] \quad \left[ \frac{e^{j\omega_0 t} - e^{-j\omega_0 t}}{2j} = \sin(\omega_0 t) \right] \\
 &= \frac{A}{\pi f} \sin(\pi f \tau) \quad \frac{\sin x}{x} \rightarrow \text{sinc}(x) \\
 &= A \tau \text{sinc}(\pi f \tau)
 \end{aligned}$$

use 

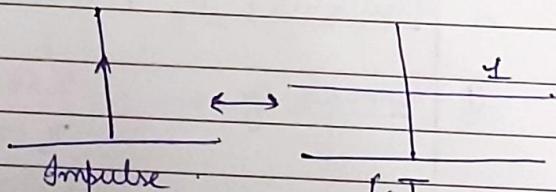
Ex 4

Fourier Transform of impulse.

$$F\{s(t)\} = \int_{-\infty}^{\infty} s(t) e^{-j2\pi f t} dt \quad \left[ \begin{array}{l} s(t) = 0 \\ \text{when } t \neq 0 \end{array} \right]$$

At  $t = 0$

$$F\{s(t)\} = 1$$



Ex 5

$$g(t) = \cos(\omega_0 t)$$

$$F\{g(t)\} = \int_{-\infty}^{\infty} \cos(\omega_0 t) e^{-j2\pi f t} dt$$

$$Ex-6. \quad g(t) = k \cdot e^{-j2\pi f t}$$

$$F\{g(t)\} = \int_{-\infty}^{\infty} k \cdot e^{-j2\pi f t} dt$$

$$= \frac{k}{j2\pi f} \left[ e^{j2\pi f t} \right]_{-\infty}^{\infty} = \frac{k}{j2\pi f} [e^{j2\pi f \infty} - 0]$$

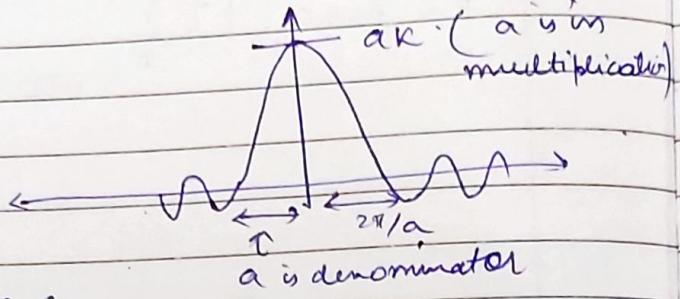
$$F\{g(t)\} = \lim_{a \rightarrow \infty} \int_{-a}^a k e^{-j2\pi f t} dt$$

$$= \lim_{a \rightarrow \infty} ak \operatorname{sinc}(fa)$$

If  $a \uparrow$   
amplitude  $\uparrow$   
and width  $\downarrow$

As  $a \rightarrow \infty$

It will become impulse.



### Properties of Fourier Transform

① Linearity:  $x_1(t) \longleftrightarrow X_1(f)$   
 $x_2(t) \longleftrightarrow X_2(f)$   
 $\text{So } a x_1(t) + b x_2(t) \longleftrightarrow a X_1(f) + b X_2(f)$ .

② Time Shifting:  $x(t) \longleftrightarrow X(f)$   
 $x(t - t_0) \longleftrightarrow X(f) e^{-j2\pi f t_0}$

③ Frequency Shifting:  $x(t) e^{j2\pi f_0 t} \longleftrightarrow X(f - f_0)$

(4) Scaling

$$x(t) \longleftrightarrow X(f)$$

$$x(at) \longleftrightarrow \frac{1}{|a|} X(f/a)$$

(5) Time Reversal

$$x(t) \longleftrightarrow X(f)$$

$$x(-t) \longleftrightarrow X(-f)$$

(6) Differentiation in Time Domain

$$n(t) \longleftrightarrow X(f)$$

$$\frac{d}{dt} x(t) \longleftrightarrow j2\pi f X(f)$$

$$\frac{d^2}{dt^2} x(t) \longleftrightarrow (j2\pi f)^2 \cancel{X(f)} X(f)$$

Ex(2)

$$g(t) = e^{-at+t}$$

$$F\{g(t)\} = \int_{-\infty}^{\infty} e^{-at+t} e^{-j2\pi ft} dt$$

$$= \int_{-\infty}^{0} e^{a(a-j2\pi f)t} dt + \int_0^{\infty} e^{(-a+j2\pi f)t} dt$$

$$= \frac{e^{(a-j2\pi f)t}}{a-j2\pi f} \Big|_{-\infty}^0 + \frac{e^{(-a+j2\pi f)t}}{-a+j2\pi f} \Big|_0^{\infty}$$

$$= \frac{1}{a-j2\pi f} + \frac{1}{a+2j\pi f}$$

$$= \frac{2a}{a^2 + 4\pi^2 f^2}$$



21-01-21 Q signal  $-A$  to  $+A$  (Peak to peak signal)

$\Rightarrow$  Signal to Noise Ratio  $\rightarrow \frac{\text{Signal Power}}{\text{Noise Power}}$

Signal to D. Noise Ratio  $\rightarrow ?$

$$\text{Step size} = \frac{2A}{m}$$

$$\text{Signal Power} = A^2/2. \quad (\text{From Area})$$

$$\overline{P^2} = \frac{s^2}{12} = \frac{A^2}{3m^2}$$

m: No. of steps.

$$\text{Signal to D. Noise Ratio} = \frac{A^2/2}{A^2/3m^2} = \frac{3m^2}{2}$$

$$(\text{SQNR})_{\text{dB}} = 10 \log (\text{SQNR})$$

$$= 10 \log \frac{3}{2} + 10 \log m^2$$

$$= 1.7 + 20 \log m$$

PAM, PWM, PPM

Pulse modulation

Process where one parameter of carrier signal varies with modulating signal.

Modulating signal  $\rightarrow$  low freq signal which want to transfer.

carrier signal  $\rightarrow$  High freq.  
(Amp or freq or Phase)

Let modulating signal  $\rightarrow 3 \text{ kHz}$

Size of Antenna  $= \lambda/4$

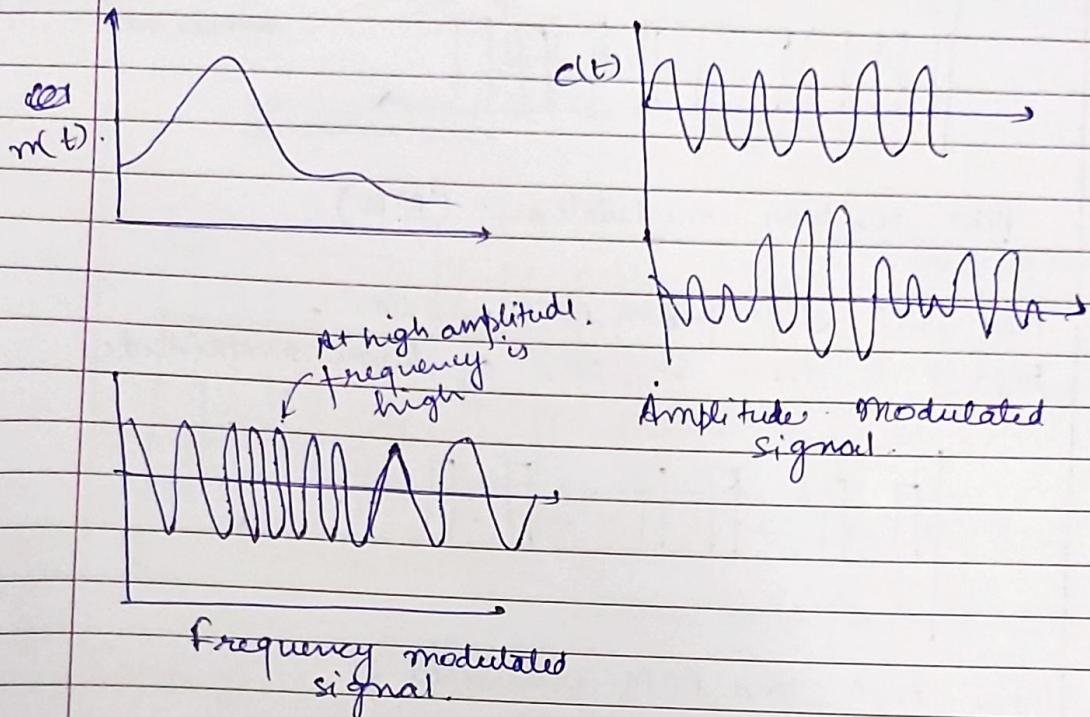
$$\lambda = c/f = \frac{3 \times 10^8}{3 \times 10^3} = 10^5$$

$$= \frac{10^5}{4} = 25 \times 10^3 \\ = 25 \text{ KM.}$$

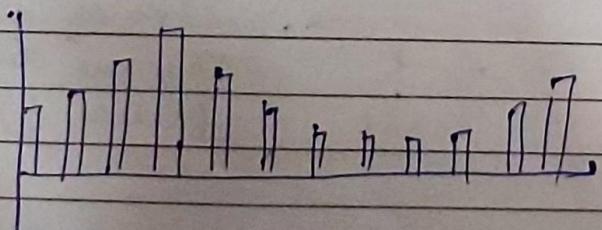
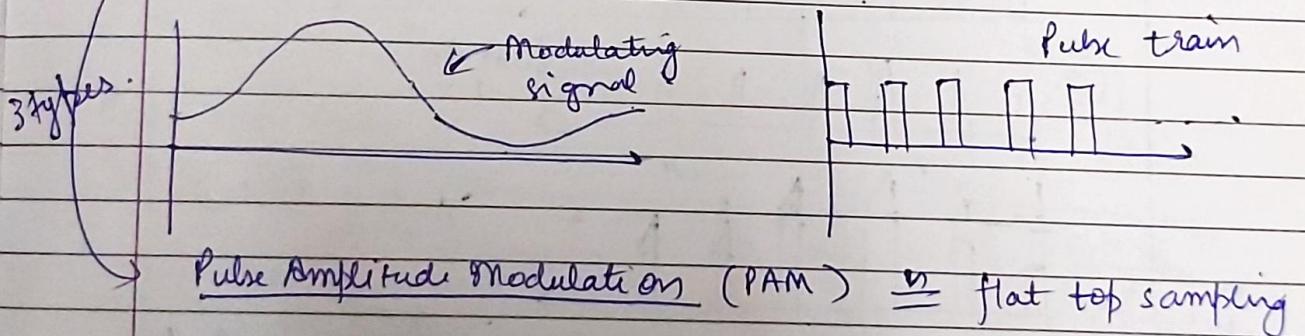
All diagrams are in time-domain

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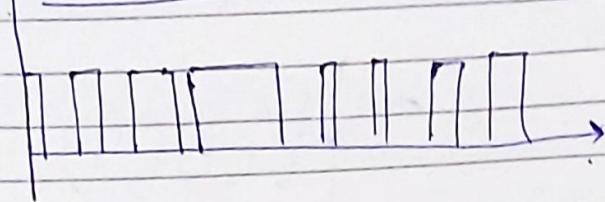
So, we have to ↑ the frequency to decrease the size of antenna.



If we modulate signal with pulse train i.e. is called pulse modulation.



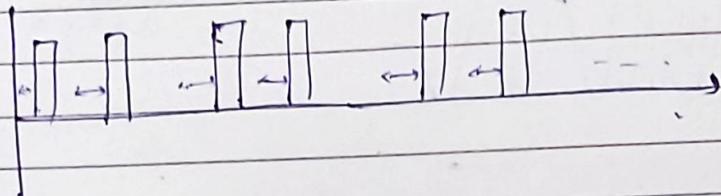
Pulse width modulation (PWM)  
 Pulse Duration n (PDW)



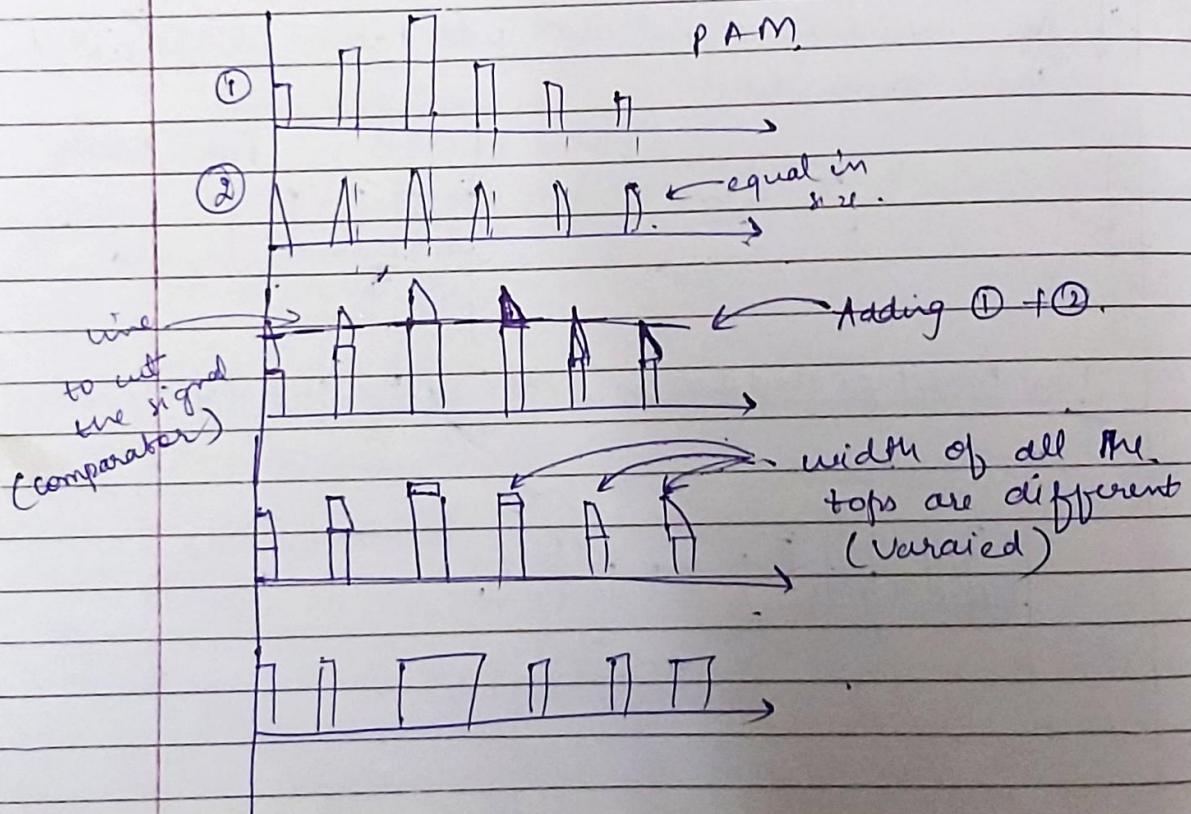
Amplitude = constant  
 width varies  
 according to  
 amplitude.

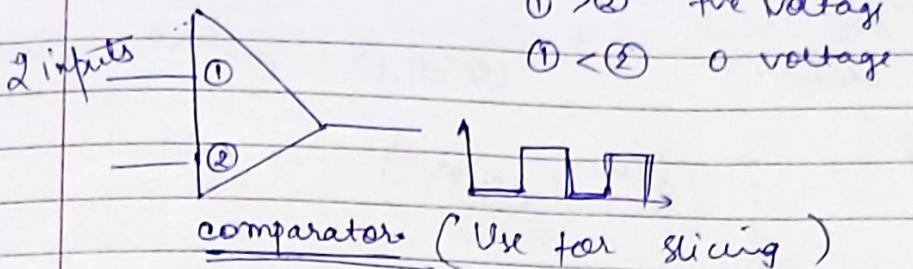
Pulse position modulation (PPM)

Pulse will shift from its position.  
 shifting is more in case of high amplitude



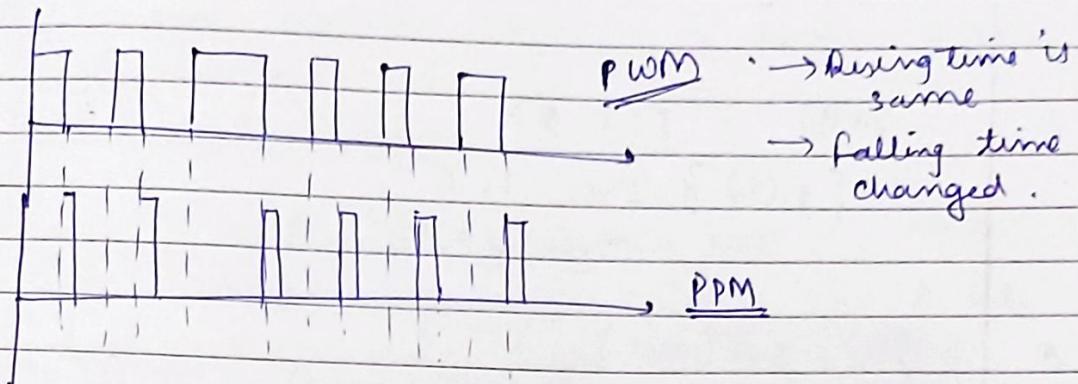
How PWM and PPM generated?





① > ② true voltage

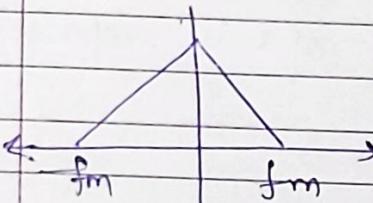
① < ② 0 voltage



→ Mono stable / Bi stable vibrator

### Sampling

Nyquist rate  $\geq 2f_m$



Exist near zero

↪ we can say low pass signal.

If situation is ↗

150 kHz    200 kHz

$f_s \geq 2f_m$

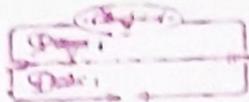
we require 400 KHz

↪ Not a nice choice

→ Large data is generated with low signal variants if we are using high sampling frequency.

### Band pass sampling

$$f_p = ?$$



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D.  $g(t) = \cos(\omega_0 t)$        $\cos(\omega_0 t) = \frac{e^{j\omega_0 t} + e^{-j\omega_0 t}}{2}$

$$\mathcal{F}\{g(t)\} = \int_{-\infty}^{\infty} \left( \frac{e^{j\omega_0 t} + e^{-j\omega_0 t}}{2} \right) e^{-j\omega t} dt$$

$$\mathcal{F}\{g(t)\} = \pi (\delta(\omega - \omega_0) + \delta(\omega + \omega_0))$$

Q.  $g_1(t) = \delta(t-3)$

$$\mathcal{F}\{g_1(t)\} = 1 \cdot e^{-j\omega \times 3}$$

$$= e^{-j3\omega}$$

Q.  $g(t) = \sin(\omega_0 t)$

$$\mathcal{F}\{g(t)\} = \pi(\delta(\omega + \omega_0) - \pi(\delta(\omega - \omega_0))$$

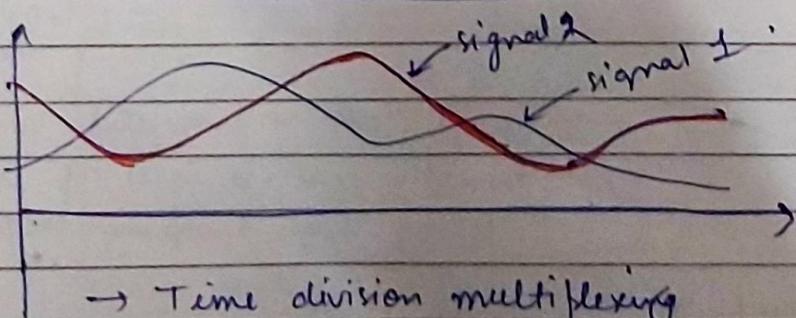
D.  $g(t) = e^{-j\omega_0 t}$

$$1 \longleftrightarrow \delta(f) \quad (\text{F.T. of } \delta \text{ is impulse})$$

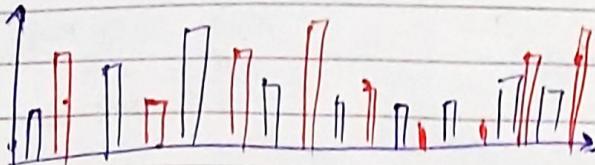
$$1. e^{j\omega_0 t} \longrightarrow \delta(\omega - \omega_0)$$

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

→ signum function.



T<sub>1</sub> system :- 24 users data can be multiplexed.



- Taking sample at alternate frequency
- To send multiple data using simple same resources.

About T<sub>1</sub> systems       $f_m = 3.4 \text{ kHz}$ .

$$f_s = 8 \text{ kHz}.$$

Quantization level = 256

Every sample values are encoded in 8 bits

- 8 bit  $\times$  24 users = 192 bits
- 1 bit for frame synchronization.

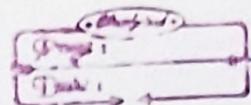
Total bits = 193 bits/frame

Data rate = 193 bits  $\times$  8 kHz.

$$= 1.544 \text{ mbps}.$$

(contd)

25/01/21



## Modulation

### Analog Modulation

- Types
  - AM, P
  - FM,
  - PM.

### Digital

#### Modulation

- Types
  - ASK. (Amp shift key)
    - Binary
    - Quadrature
  - PSK.
    - M-ary
  - FSK.
    - Q.
    - M-ary

### Pulse

- Types
  - PAM
  - PPM
  - PWM

PCM (Pulse code Modulation)

DPCM (Differential " " " )

DM (Delta Modulation)

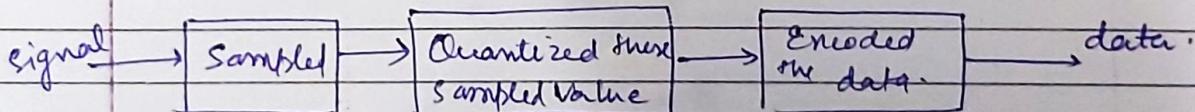
ADM (Adaptive Delta Modulation)

These techniques

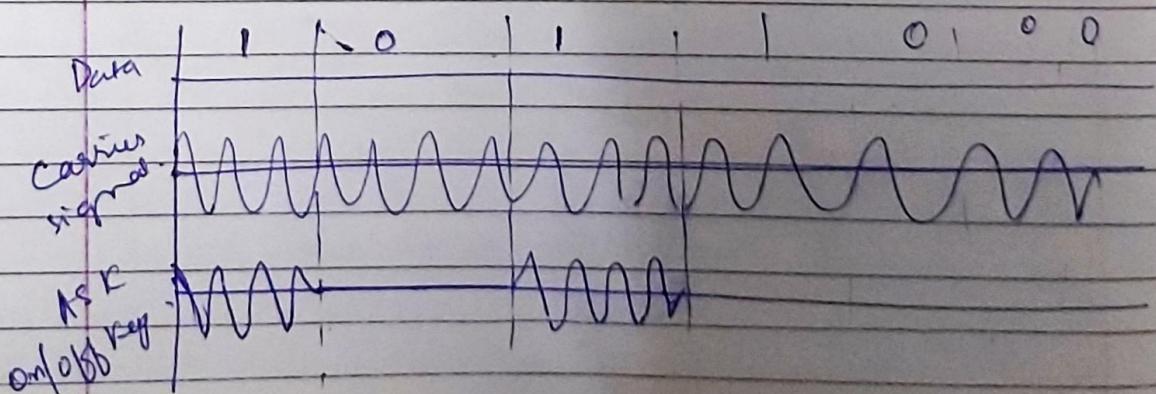
helps → Analog  $\xrightarrow{\text{to}}$  Digital Data form.  
convert

These are  
not modulation  
techniques.

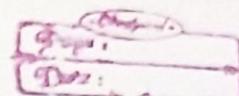
Now we have data that to be modulated not signal.



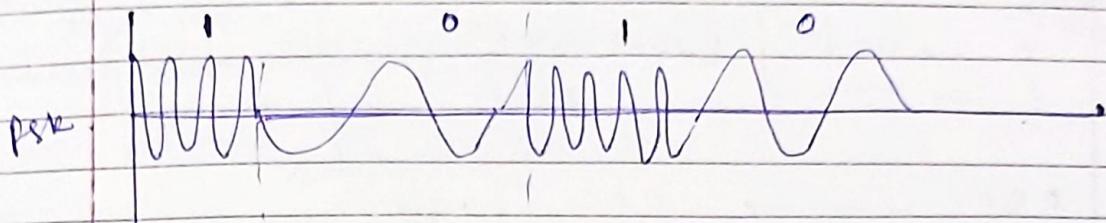
This data is need to be transmitted from one to another end



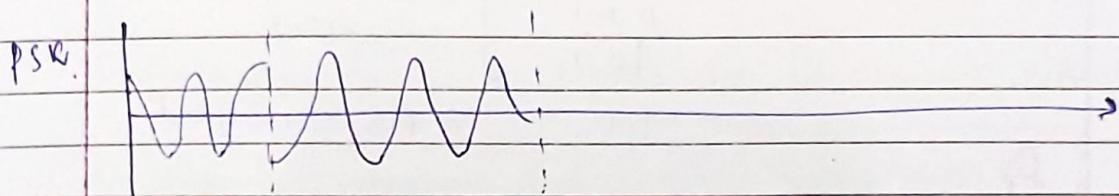
→ bit Duration  
→ symbol Duration.  $f = n f_c$



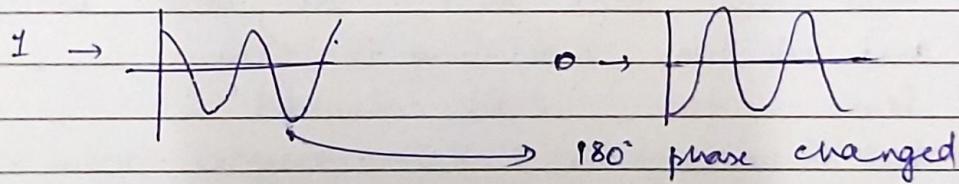
ASK → for '1' data, Signals ~~are~~ <sup>carrier wave is</sup> transmitted  
"0", No signal is transmitted



PSK → for '1' data → high frequency.  
For '0', → low frequency.

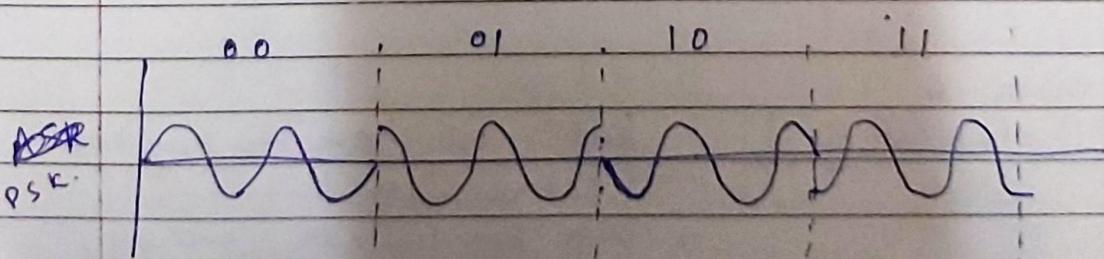


PSK → For '1' data, with one phase -  
for '0' data, with opposite phase.



Type → Binary → Single bit is modulated at a time  
→ Quadrature → 2 bits  
many → General ( $m = 8, 16, 32, \dots$ )

For Quadrature (4 combinations are possible)



Phase Diff =  $\pi/2$  (for 4 states)

→ No of levels ↑ Probability of error also ↑.

Example

0 → 000 ] Probability of error = ?  
 1 → 111 ] We are considering  
 0 → 000  
 1 → 111

000 → 001 ] 1 bit corrupted.  
 010  
 100

110 ] 2 bits corrupted.  
 011  
 101

111 3 bits corrupted.

Probability of error =  $\frac{1}{n}$  data

- When '1 bit corrupted' is received, receiver will detect that there is an error and ask for re-sending the data.
  - Same for '2 bit corrupted'.
  - But when all 3 bits corrupted now receiver can't identify that it is a corrupted data or not.
- so, probability of error =  $\frac{1}{8}$ .

2) 0 → 0000 ] Probability of error =  $\frac{1}{16}$   
 1 → 1111

~~extra data~~

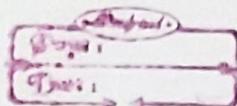
Probability of error but we have to pay for extra data.

→ Bandwidth, energy,

→ for improving system buffer size etc.  
 we require lot of extra things.

Given  $A \cos(2\pi f_c t)$  Power =  $A^2/2$

$$\frac{E}{T} = P \quad \frac{E}{T} = \frac{A^2}{2} \Rightarrow A = \sqrt{\frac{2E}{T}}$$



→ so, we have some limitations for adding extra data.

~~Topic 1/21  
97~~

### Binary Phase Shift Keying (BPSK)

$$S_1(t) = \sqrt{\frac{2E_b}{T_b}} \cos(2\pi f_c t)$$

$$S_2(t) = -\sqrt{\frac{2E_b}{T_b}} \cos(2\pi f_c t + \pi) = -\cos\left(\frac{2E_b}{T_b} \cos(2\pi f_c t)\right)$$

where  $E_b \rightarrow$  Bit energy  $T_b \rightarrow$  Bit duration

Orthogonal Signal :- signals lie in different directions  
→ interference is minimum.

Orthonormal function  $\Rightarrow \int_T \phi_i(t) \phi_j(t) dt = 0$

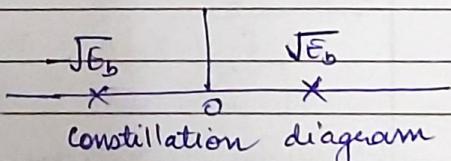
$\phi_2$        $\phi_1$        $\phi_i(t) = \sqrt{\frac{2}{T_b}} \cos(2\pi f_c t)$

$$S_1(t) = \sqrt{E_b} \phi_1(t)$$

$$S_2(t) = -\sqrt{E_b} \phi_1(t)$$

$S_1$  and  $S_2$  are represented using one orthogonal function. which means both signal lies in one dimension.

$$S_{11} = \int_0^{T_b} S_1(t) \phi_1(t) dt$$



$$= \int_0^{T_b} \sqrt{E_b} \phi_1(t) \cdot \phi_1(t) dt$$

$$= \sqrt{E_b} \int_0^{T_b} \phi_1(t) \cdot \phi_1(t) dt$$

$$|S_{11}| = \sqrt{E_b}$$

$$S_{22} = S_{11} = \int_0^T S_2(t) \phi_1(t) dt$$

$$= -\sqrt{E_b}$$

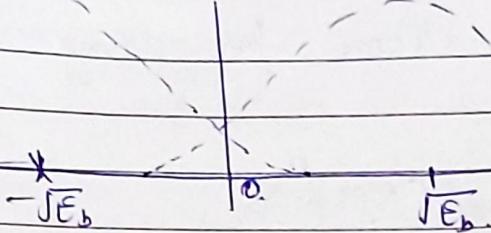
orthonormal

$$\int_T \phi_i(t) \phi_j(t) dt = 0$$

$$i = j$$

$$\int_T \phi_i(t) \phi_j(t) dt = 1$$

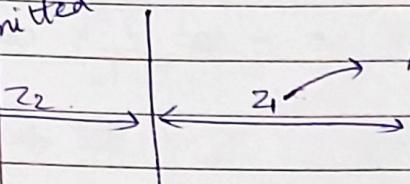
$$i \neq j$$



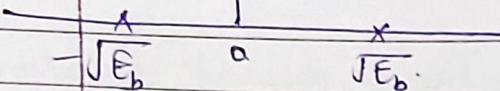
constellation diagram.

$s_1, s_2 \rightarrow$  coordinates of signal space

0 was transmitted



After detection if any values in this region, it means 1 was transmitted



→ Only one signal is transmitted at a time through the channel.

→ If received signal is  $x(t)$

$$\text{Observation scalar} \Rightarrow x_i = \int_{-\infty}^{T_b} x(t) \phi_i(t) dt$$

Acc to  $x_i$ , we will decide 0 is transmitted or 1

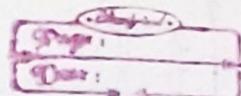
Error probability (calculation)

$$f_{x_i}(x_i | 0) = \frac{1}{\sqrt{\pi N_0}} \exp \left[ -\frac{1}{N_0} (x_i - s_1)^2 \right]$$

likelihood function when signal 0 was transmitted

$$f_{x_i}(x_i | 0) = \frac{1}{\sqrt{\pi N_0}} \exp \left[ -\frac{1}{N_0} (x_i + \sqrt{E_b})^2 \right]$$

$$\text{Prob of Error (0)} = \int_0^\infty f_{x_i}(x_i | 0) dx_i = \frac{1}{2}$$



$$= \int_0^{\infty} \frac{1}{\sqrt{\pi N_0}} \exp \left[ -\frac{1}{N_0} (x_1 + \sqrt{E_b})^2 \right]$$

let  $\frac{1}{\sqrt{N_0}} (x_1 + \sqrt{E_b}) = z$

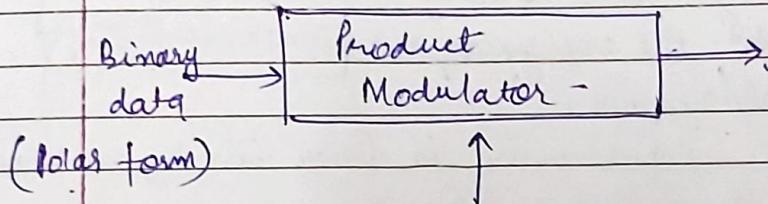
$$P_e(0) = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} \exp(-z^2) dz.$$

$$P_e(0) = \frac{1}{2} \operatorname{erfc} \left( \sqrt{\frac{E_b}{N_0}} \right)$$

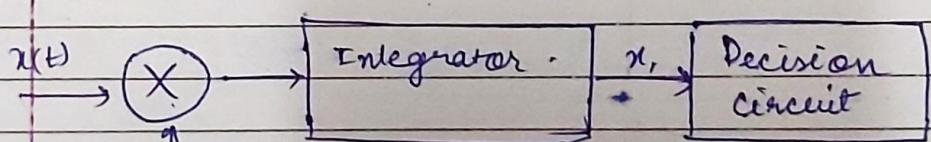
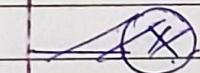
$$\operatorname{erfc}(u) = \frac{2}{\sqrt{\pi}} \int_u^{\infty} e^{-r^2} dr.$$

$$P_e(1) = \frac{1}{2} \operatorname{erfc} \left( \sqrt{\frac{E_b}{N_0}} \right)$$

$$P_{e+} = \frac{1}{2} (P_e(0) + P_e(1))$$



$$\phi_1(t) = \int \frac{2}{T_b} \cos(2\pi f_c t)$$



$$\phi_i(t)$$

0 if  $x_i < 0$       1 if  $x_i > 0$



uniform Distribution

28/01/21

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For any binary modulation we require two functions.

$$S_1(t) = \begin{cases} S_1(t) & 0 \leq t \leq T \\ S_2(t) & 0 \leq t \leq T \end{cases} \Rightarrow 0 \quad 1$$

$$y(t) = s_i(t) \otimes h(t) + n(t)$$

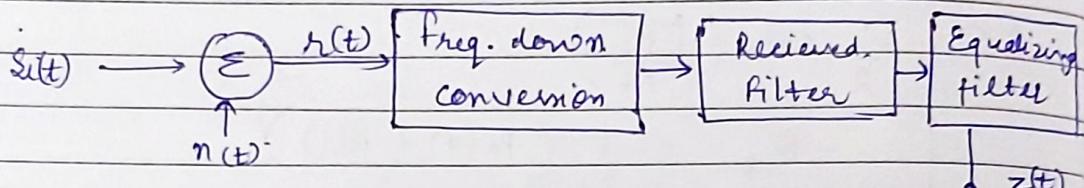
~~Received signal~~

↳ impulse response of channel.

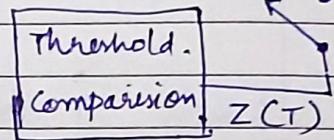
for ideal distortionless channel

$$y(t) = s_i(t) + n(t) \quad i = 1 \text{ or } 2.$$

↳ noise

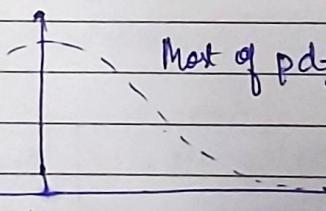


→  $z(t)$  has a voltage value directly proportional to the energy of the received signal/symbol.



- # We consider that input noise is Gaussian by nature.
- # Input noise is Gaussian random process.
- # Received filter is a linear

Gaussian.



Most of pdf's are Gaussian.

$$f_x(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

↳ lies near its mean value

$$z(t) = a_i(t) + n_o(t)$$

It will also be gaussian

$$z = a_i + n_o$$

$z(t)$  is also a Gaussian random variable with a mean either  $a_1$  or  $a_2$ .

$$P(n_0) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(n_0 - \mu_0)^2}{2\sigma^2}}$$

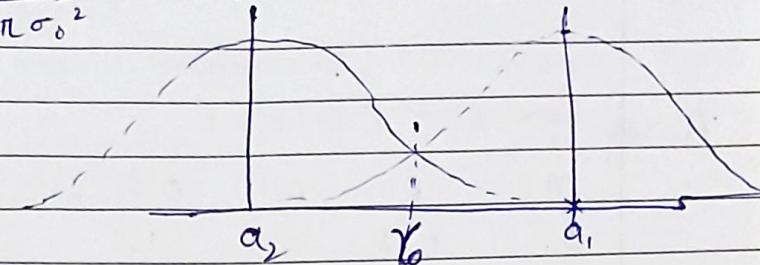
let mean of  $n_0(t) = 0$

$$= \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2} \frac{(n_0)^2}{\sigma^2}}$$

$$P(z|s_1) = \frac{1}{\sqrt{2\pi\sigma_1^2}} e^{-\frac{1}{2} \frac{(z - a_1)^2}{\sigma_1^2}}$$

$$P(z|s_2) = \frac{1}{\sqrt{2\pi\sigma_2^2}} e^{-\frac{1}{2} \frac{(z - a_2)^2}{\sigma_2^2}}$$

Diffr  $a_1 - a_2$ . ↑  
error ↓



Where should be  $\gamma_0$ ?

$P(s_1) = P(s_2)$  Probability of  $s_1$  = Prob. of  $s_2$

The likelihood ratio.

$$\frac{P(z|s_1)}{P(z|s_2)} \stackrel{H_1}{\geq} \frac{P(s_2)}{P(s_1)} \quad \text{Bay's Theorem}$$

Due to this

$$z \stackrel{H_1}{\geq} \frac{a_1 + a_2}{2}$$

$\gamma_0$

$$P(e|s_1) = \int_{-\infty}^{\gamma_0} P(z|s_1) dz$$

$$P(e|s_2) = \int_{\gamma_0}^{\infty} P(z|s_2) dz$$

$$P_b (\text{Probability of bit error}) = P(e|s_1) (P(s_1) + P(e|s_2), P(s_2))$$

$$P_b = \frac{1}{2} [P(e|s_1) + P(e|s_2)]$$

$$P(e|s_1) = P(e|s_2)$$

$$P(h_2|s_1) = P(h_1|s_2)$$

$$\text{Probability of error} = \int_{y_1}^{y_2} P(z|s_2) dz = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(z-a_2)^2}{2\sigma^2}}$$

Q1/12 for Practical → MATLAB, // clear a

Commands (On command window) some commands // clear all

Array →  $n = 0:1/10:1$  // help - - -

$n = 0.0 \quad 0.1 \quad 0.2 \quad 0.3 \quad 0.4 \quad 0.5 \quad 0.6 \quad 0.7$   
 $0.8 \quad 0.9 \quad 1.0$

Array →  $A = [1 \ 2 \ 4 \ 7 \ 9, 2]$  Both are valid  
space → commas →

$A = 1 \ 2 \ 4 \ 7 \ 9 \ 2$ .

→  $A = [1 \ 2 \ 3; 2 \ -1 \ 0; 0 \ 3 \ 5]$  2-D array  
3x3 Matrix

→  $A + B$

Both are matrices

→  $A - B$

→  $C = A + 2$ .

→  $A * B$

→  $A'$  (Transpose of A)

→  $A + 2$ .

→ for writing program we have to open script window

→ initiate your program with clc and clear all.

-  $clc;$   $\text{clear all};$

-  $t = 0:1/100:1$

-  $f = 2;$

-  $x = \cos(2 * \pi * f * t);$

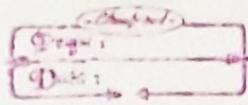
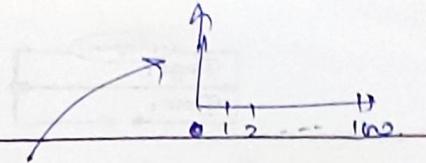
-  $\text{Plot}(x)$  → y variable on I

plot ( $t, x$ ) (Case II.)

? .  
y-var

y-var

Example of  
Programs



case I.  $n$  - contains index of array starting with 1

case II  $n$  - contain values of  $t$ .

for running more than 1 plots simultaneously

- subplot (2, 1, 1) position.  
 ↗ columns  
 ↘ row.

→ for array operations we have use . before operators.

plot ( $t$ ,  $x_1$ );

$f_2 = 10$ ;

$x_2 = \cos(2 * \pi * f_2 * t)$ ;

- subplot (2, 1, 2)

- plot ( $t$ ,  $x_2$ )

eg :- . +, . ^, . /  
 ↗ Power operator

### Loop

for  $n = 1 : 2 : 30$ ;

$a = 2^n$ ;

end

if - else

if

else

end

for  $i = 1 : 2 : 30$ ;

$a(i) = 2^i$ ;

end.

### Random variable

$a = \text{Rand}(10) \rightarrow$  generate  $10 \times 10$  matrix

$a = \text{Rand}(1, 10) \rightarrow$  generate  ~~$10 \times 1$~~   $1 \times 10$

↳ Uniform Random variable (i.e. from 0 to 1)

$b = \text{Randn}(1, 10)$

↳ Gaussian random variables

$\text{hist}(a)$

$\text{hist}(b)$

Date  
01-02-21

(Classmate)  
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## Binary frequency Shift Keying (BFSK)

$$S_1(t) = \sqrt{\frac{2E_b}{T_b}} \cos(2\pi f_1 t) \quad S_2(t) = \sqrt{\frac{2E_b}{T_b}} \cos(2\pi f_2 t)$$

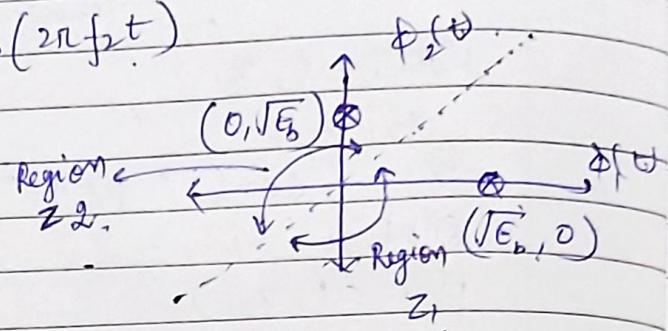
the orthonormal basis function.

$$\phi_1(t) = \sqrt{\frac{2}{T_b}} \cos(2\pi f_1 t)$$

$$\phi_2(t) = \sqrt{\frac{2}{T_b}} \cos(2\pi f_2 t)$$

$$S_1(t) = \sqrt{E_b} \phi_1(t)$$

$$S_2(t) = \sqrt{E_b} \phi_2(t)$$



$$S_{11} = \int_0^{T_b} S_1(t) \cdot \phi_1(t) dt = \sqrt{E_b}$$

$$\Rightarrow S_{12} = \int_0^{T_b} \sqrt{E_b} \phi_1(t) \phi_2(t) dt = 0.$$

$$S_{22} = \int_0^{T_b} \sqrt{E_b} \phi_2(t) \phi_2(t) dt = \sqrt{E_b}$$

Middle point should be boundary.

→ Error in BFSK is more than BPSK because of less distance between points  $(0, \sqrt{E_b})$  and  $(\sqrt{E_b}, 0)$

If the received signal is  $x(t)$ .

Observations

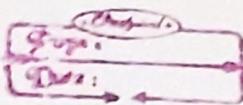
$$x_1 = \int_0^{T_b} x(t) \phi_1(t) dt$$

scalar

$$x_2 = \int_0^{T_b} x(t) \phi_2(t) dt$$

when  $x_1 > x_2 \Rightarrow$  Region  $Z_1$

$x_1 < x_2 \Rightarrow$  Region  $Z_2$



⇒ define a new random variable  $L$  whose sample value  $l$  is

$$l = x_1 - x_2.$$

if  $l > 0 \Rightarrow$  Region 2<sub>1</sub>       $l < 0 \Rightarrow$  Region 2<sub>2</sub>

Expectation value

$$E[L/\pm] = E[x_1/\pm] - E[x_2/\pm]$$

$$E[l/\pm] = \sqrt{E_b} - 0 = \sqrt{E_b}$$

$$E[L/0] = E[x_1/0] - E[x_2/1]$$

$$E[l/0] = 0 - \sqrt{E_b} = -\sqrt{E_b}$$

$$V[l] = \text{Var}[x_1] + \text{Var}[x_2] = N_b/2 + N_o/2 = N_o$$

$L \rightarrow$  mean = 0, Variance =  $N_o$

⇒ The conditional prob. of random var.  $L$  when

0 was transmitted

$$f_L(l/0) = \frac{1}{2\pi N_o} \exp\left[-\frac{(l + \sqrt{E_b})^2}{2N_o}\right]$$

$$P(l/0) = P(l > 0 \mid \text{symbol 0 was transmitted})$$

$$= \int_{-\infty}^{\infty} f_L(l/0) dl$$

$$= \frac{1}{\sqrt{2\pi N_o}} \int_{-\infty}^{\infty} \exp\left[-\frac{(l + \sqrt{E_b})^2}{2N_o}\right] dl$$

Let

$$\frac{l + \sqrt{E_b}}{\sqrt{2N_o}} = z$$

$$P(l/0) = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{\infty} e^{-z^2} dz.$$

$$P(l/0) = \frac{1}{2} \operatorname{erfc}\left(\sqrt{\frac{E_b}{2N_o}}\right)$$

$$\text{Similarly, } P(e|z) = \frac{1}{2} \operatorname{erfc}\left(\sqrt{\frac{E_b}{2N_0}}\right)$$

$$P_e = \frac{1}{2} [P(e|0) + P(e|1)]$$

$$P_e = \frac{1}{2} \operatorname{erfc}\left(\sqrt{\frac{E_b}{2N_0}}\right)$$

Error is more in case of BFSK.

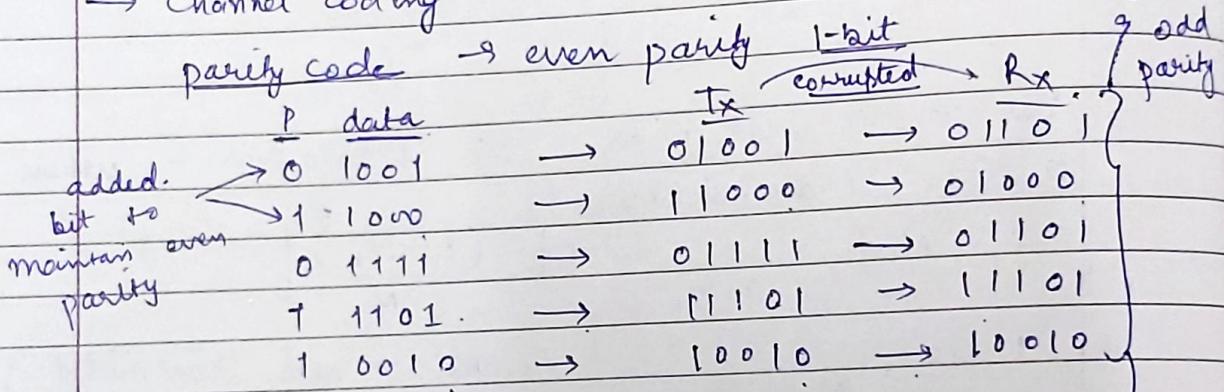
Home Assign :- Binary Amplitude Shift Keying or.  
Binary ON-OFF Keying

### Error Correcting codes

→ source coding

→ channel coding

to minimize effect of noise on channel.



→ If 2 bit corrupted then we can't identify if it is corrupted or not

so, we can say this is wrong data

### Linear Block code ( $n, k$ )

$n \rightarrow$  No. of coded bits

$k \rightarrow$  No. of data bits

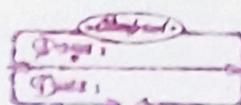
$n-k \rightarrow$  No. of Parity bits

$$\bar{C} = \bar{m} G$$

$m \rightarrow$  message array  
(data array)  $b^k$

$G$  - Generator matrix

## Hamming code (7, 4)



$$\bar{c} = [P : m]$$

$$m = [1 \ 0 \ 1 \ 0]$$

Generator  
Matrix

$$G = [P^T : I_k]$$

$$n \rightarrow 7$$

$$k \rightarrow 4$$

$$n-k \rightarrow 3$$

$$\bar{G} = \begin{bmatrix} 1 & 1 & 0 & : & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & : & 0 & 1 & 0 & 0 \\ 1 & 1 & 1 & : & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & : & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\bar{c} = [1 \ 0 \ 1 \ 0] \bar{G}$$

$$c = [0 \ 0 \ 1 : 1 \ 0 \ 1 \ 0]$$

$$\boxed{c_{i+j} = m_i G_i + m_j G_j} \rightarrow \text{Property of code}$$

$$= (m_i + m_j) G_i$$

$$\bar{H} = [I_{n-k} : P^T] \quad H - \text{parity check matrix}$$

$$H G^T = [I_{n-k} : P^T] \begin{bmatrix} P^T \\ I_k \end{bmatrix} = P^T + P^T = 0$$

$$c = m G.$$

$$c H^T = m G H^T$$

$$\boxed{c H^T = 0.}$$

$$c = 1 \ 0 \ 1 \ 0 \ 0 \ 1 \ 1$$

Received code  
vector

$$r = c + e \quad \hookrightarrow \text{error vector}$$

$$e = 0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0$$

$$S = r H^T$$

$$r = 1 \ 1 \ \boxed{1} \ 0 \ 0 \ 1 \ 1$$

↳ error bit

$$S = (c + e) H^T$$

$$= c H^T + e H^T$$

$$\boxed{S = e H^T} \rightarrow 0.$$

Syndrome :- Gives info about error location.

If  $S = 0$  then there is no error

## Hamming Distance

↳ No. of elements which are differ from one code to another.

Eg:-  $C_i = 1011001$   
 $C_j = 1101001$

H. Distance = 2

$d_{min} \rightarrow$  minimum distance.

→ An  $(n, k)$  linear block code of minimum distance  $d_{min}$  can correct upto 't' errors errors if and only if  $t \leq \left\lfloor \frac{1}{2} (d_{min} - 1) \right\rfloor$

For.  $(7, 4)$   $d_{min} = 3 \Rightarrow t \leq 1$

→ We can only correct single bit error in  $(7, 4)$  codeword.

Syndrome	Error Pattern.
000	0000000 (No error)
100	1000000 (1 <sup>st</sup> bit corrupted)
010	0100000 (2 <sup>nd</sup> bit)
001	0010000
110	0001000
011	0000100
111	0000010
101	0000001

Eg:- received code vector  $r = [11010010]$

Find the correct transmitted code word.

$$S = rH^T$$

$$S = [11010010] H^T$$

$$\therefore S = 001$$

$$H^T = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ \vdots & \vdots & \vdots \\ m-k & & p_1 \end{bmatrix}$$

$$S = 001 \quad e = 0010000$$

→ 3<sup>rd</sup> bit is corrupted.

$$\begin{aligned} r &= 1100010 \\ &= 1110010 \end{aligned}$$

Prop. of syndrome

- # The syndrome depends only on the error pattern and not on the transmitted code word.

$$S = RH^T = (C + e)H^T$$

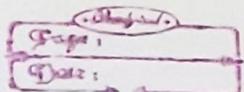
$$S = eH^T$$

- # All error patterns that differ at most by a code word have the same syndrome

- # The syndrome is the sum of those columns of the matrix  $H$  corresponding to the error location.

$$\# t \leq \left\lceil \frac{1}{2} (d_{\min} - 1) \right\rceil \quad \underline{\text{Proof}} = ?$$

Hamming weight



$s = 001 \quad e = . 0010000$   
 $\rightarrow 3^{\text{rd}}$  bit is corrupted.

$$\begin{aligned} s + e &= . 1100010 \\ &= 1110010 \end{aligned}$$

Prop. of syndrome:

- # The syndrome depends only on the error pattern and not on the transmitted code word.

$$s = rH^T = (c+e)H^T$$

$$s = eH^T$$

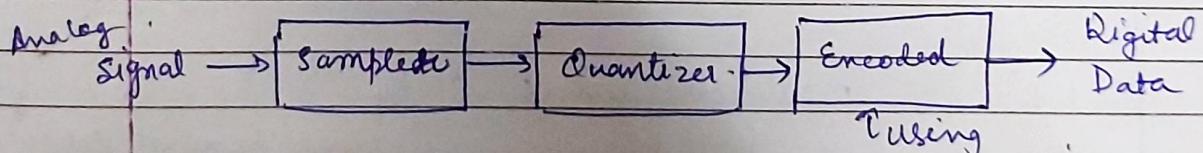
- # All error patterns that differ at most by a code word have the same syndrome

- # The syndrome is the sum of those columns of the matrix  $H$  corresponding to the error location.

$$t \leq \left\lceil \frac{1}{2} (d_{\min} - 1) \right\rceil \quad \underline{\text{Root}} = ?$$

Hamming weight

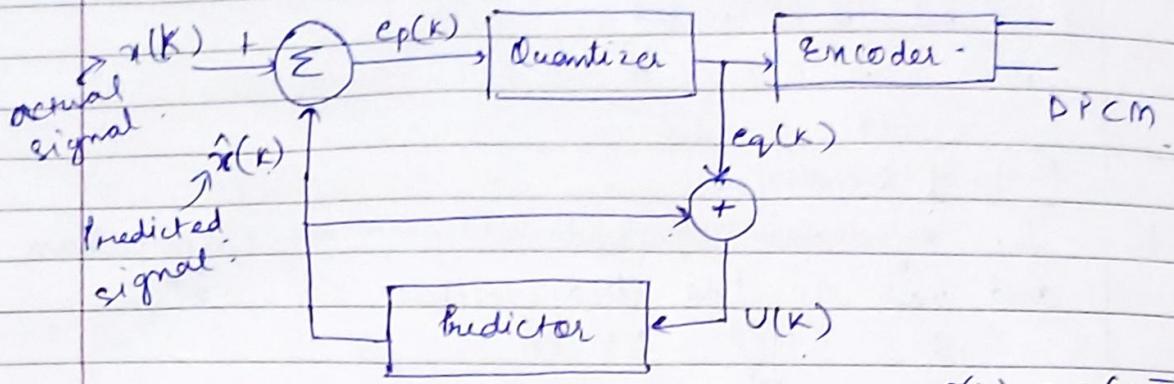
## 06-2-21 Pulse Code Modulation (PCM)



$\rightarrow$  DPCM Problem:- large data size.      Source coding

for reducing data size instead of sending the exact sampled value we are sending the difference b/w two consecutive sampled value.

## Differential PCM



$$e_p(k) = x(k) - \hat{x}(k)$$

→ Prediction error of  $k^{\text{th}}$  sample

$e_q(k)$  = Quantized prediction error

$q(k)$  = Quantization error

$$e_q(k) = e_p(k) + q(k)$$

I/P to the predictor

$$u(k) = \hat{x}(k) + e_q(k)$$

$$u(k) = \hat{x}(k) + e_p(k) + q(k)$$

$$u(k) = x(k) + q(k)$$

SQNR (signal to quantization noise ratio)

$$\text{SQNR} = \frac{\text{var of } x(k)}{\text{var of } q(k)} = \frac{\text{var}[x(k)]}{\text{var}[e_p(k)]} \times \frac{\text{var}[e_p(k)]}{\text{var}[q(k)]}$$

$$\text{SQNR.} = \text{prediction Gain} \times \text{SNR}_p$$

( $G_p$ ) ↑ signal to noise ratio.

→  $G_p$  should be as high as possible

One Tap Predictor :-  $k^{\text{th}}$  value depend on  $k-1^{\text{th}}$  value

$$\hat{x}(k|k-1) = \hat{x}(k)$$

Original  
Date:

$$\hat{x}(k) = a \cdot u(k-1) \quad u(k-1 | k-1) = u(k-1)$$

$\downarrow$

Prediction coefficient

$$e_p(k) = x(k) - \hat{x}(k)$$

$$e_p(k) = x(k) - a \cdot u(k-1)$$

Mean square error

$$E[e_p^2(k)] = E[(x(k) - a \cdot u(k-1))^2]$$

$$E[x^2(k)] = \text{signal power} = R(0)$$

$$E[-2a x(k) u(k-1)] = -2a E[x(k) \cdot u(k-1)]$$

let quantization noise is very small -

$$= -2a E[x(k) \cdot x(k-1)]$$

$$= -2a R(1)$$

↑ auto correlation

$$E[e_p^2(k)] = R(0) - 2a R(1) + (?)$$

$$(?) = E[a^2 u(k-1) u(k-1)]$$

$$= E[a^2 x(k-1) x(k-1)]$$

$$= a^2 E[x(k-1) x(k-1)]$$

$$= a^2 E[x(k) x(k)]$$

$$= a^2 R(0)$$

Auto correlation

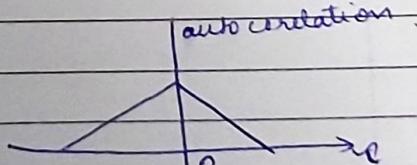
$$E[e_p^2(k)] = R_x(0) - 2a R_x(1) + a^2 R_x(0)$$

$$R(z) = E[x(t_1) x(t_1+z)]$$

$$E[e_p^2(k)] = R_p(0)$$

$$R_p(0) = R_x(0) - 2a R_x(1) + a^2 R_x(0)$$

↑  
Prediction error power  
Signal power



$$R(0) = [x(t_1) * x(t_1)]$$

= signal power

$$R_{sp}(0) = R_x(0) \left[ 1 - 2a \frac{R_x(1)}{R_x(0)} + a^2 \right]$$

This relation is used to calculate value of  $a$

$$\boxed{G_p = \frac{R_x(0)}{R_{sp}(0)}}$$

$$\boxed{G_p = \frac{1}{1 - 2a \frac{R_x(1)}{R_x(0)} + a^2}}$$

Maximize it  
Maximize it to  
get value of  $a$