

26/2/20

Bernoulli trials : $P(H) = 0.7$

$P(T) = 0.3$

Let $n=10$

$P(HH\cdots H) = (0.7)^{10}$ - seems to be event with highest probability
- but atypical event
- least possibility of occurrence

$$\rightarrow X_1, X_2, \dots, X_n \text{ & } \frac{1}{n} \sum X_i \rightarrow E(X)$$

consider $\log(pmf) = \log(p(x_1)), \dots, \log(p(x_n))$

apply LLN on $\log(p(x_i))$

$$\frac{1}{n} \sum_{i=1}^n \log(p(x_i)) \rightarrow E(\log(p(x_i))) = -H(X)$$

$$\frac{1}{n} \sum_{i=1}^n \log(p(x_i)) = \frac{1}{n} \log \left(\prod p(x_i) \right)$$

$$= \frac{1}{n} \log(p(X_1, X_2, \dots, X_n)) \rightarrow -H(X)$$

$$p(X_1, X_2, \dots, X_n) \rightarrow 2^{-nH(X)}$$

i.e., for a small ϵ

$$2^{-n(H(X)+\epsilon)} \leq p(X_1, X_2, \dots, X_n) \leq 2^{-n(H(X)-\epsilon)}$$

$$\text{at } \epsilon \rightarrow 0 : p(X_1, X_2, \dots, X_n) = 2^{-nH(X)}$$

$$H(X) = -0.7 \log(0.7) - 0.3 \log(0.3) = 0.88.$$

$$2^{-nH(X)} = 2^{-10 \times 0.88} = 2^{-8.8} = 2.24 \times 10^{-3}$$

Cardinality of such set $\rightarrow 2^{nH(X)} = 2^{8.8}$

$$|A_C| \rightarrow 445.72$$

\therefore $\neq H, ST$ will be most likely events - Typical event.

Universal Coding (Ziv-Lempel)

$X = \{a_1, a_2, \dots, a_M\}$ steady state have been reached



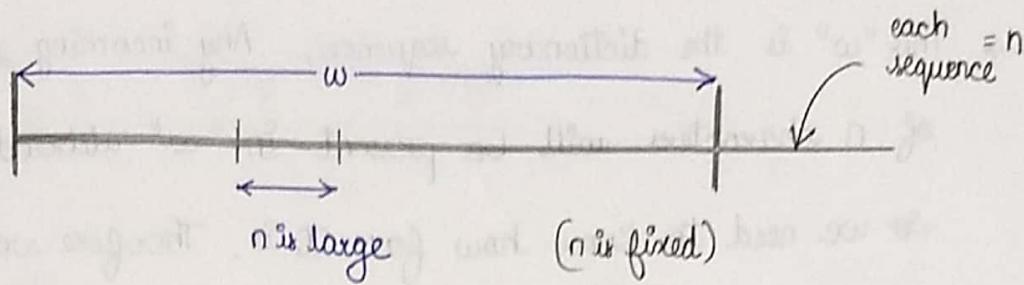
Stationary

$$H(X|S) = H_{\infty}(X)$$

normally $\lceil \log_2 M \rceil$ bits are required to represent

M symbols (a_1, a_2, \dots, a_M) \equiv trivial coding

\rightarrow For w characters taking together: $w \lceil \log_2 M \rceil$ (trivially)



$2^{-nH_{\infty}(X)}$ probability for each sequence (n symbol sequence)
 \Rightarrow there are $2^{nH_{\infty}(X)}$ such sequences.
 L each is equiprobable.

- $n \cdot 2^{nH_{\infty}(X)}$
- Sequence of characters this long
 - all $2^{nH_{\infty}(X)}$ (n symbol) sequences occur typically in this sequence.

Eg: $M = 64, \lceil \log M \rceil = 6$

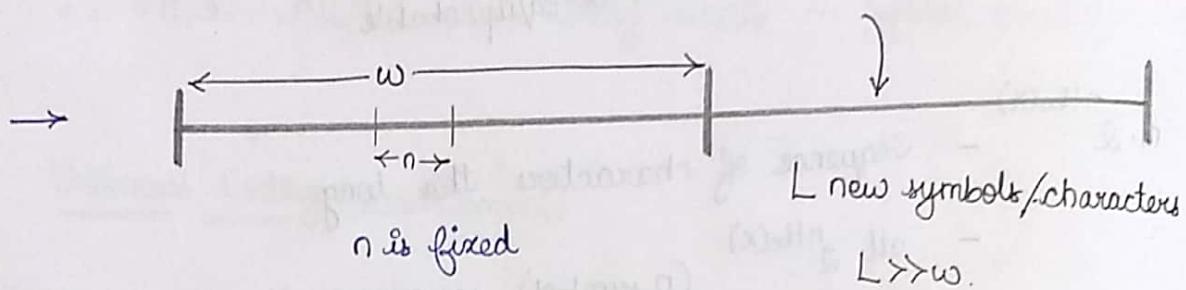
$n = 10, H_{\infty}(X) = 3$

10×2^{30} - we can see all sequences of 10 bits long, where each one is equiprobable, in this much length string.

$$w \sim 10 \times 2^{30}$$

$$\therefore w \sim n \times 2^{nH_{\infty}(X)}$$

- This " w " is the dictionary sequence. Any incoming sequence of n characters will be present in " w " already. So we need to know how far it is. Therefore we encode this distance.
- Simply we are storing / encoding the pointers of n -length sequences in " w ".



→ Do trivial coding for dictionary string (w):

$$w \lceil \log_2 M \rceil = \text{number of bits required for } w$$

→ # of new symbols = L

$$\# \text{ of } n\text{-symbol sequences} = \frac{L}{n}$$

$$\# \text{ of bits for coding distance} = \log_2 L$$

$$\frac{L}{n} \log_2 w = \text{number of bits required for } L$$

$$\rightarrow \text{Total number of bits} = (w+L) \lceil \log_2 M \rceil + \frac{L}{n} \log_2 w$$

$$\text{no. of bits per symbol} = \frac{\omega \lceil \log M \rceil + \frac{L}{n} \log \omega}{L + \omega}$$

$$\therefore L \gg \omega$$

$$\approx 0 + \frac{1}{n} \log \omega = \frac{1}{n} \log \omega$$

$$= \frac{1}{n} \log (n \cdot 2^{nH_\infty(X)})$$

$$= \frac{1}{n} [\log n + nH_\infty(X)]$$

asymptotically ($n \rightarrow \infty$)

$$\approx H_\infty(X)$$

$$\frac{1}{n} \approx \frac{0.100}{n} \quad \frac{0.11001000000}{n} = \underbrace{0.1}_{a} \underbrace{0.001}_{b}$$

27/2/20

$$\chi = \{a, b, c, d, e, f, g, h\} ; \lceil \log_2 |\chi| \rceil = 3$$

$$H_{\text{bo}}(X) = 1, \quad n = 64, \quad w \sim n \cdot 2^{H_{\text{bo}}(X)} = 4 \times 2^4 = 64.$$

• Markov chain.

$w : 64 \text{ char}$ $\xrightarrow{38}$

aacaabcaadfaadeacbabf ggaa ghadeabeabbefggahahabc...
deacdfgha |

* Trivial coding :	a - 1000	e - 1100
with one extra bit	b - 1001	f - 1101
	c - 1010	g - 1110
	d - 1011	h - 1111

$$w: \underline{1000} \quad \underline{1000} \quad \underline{1010} \quad \underline{10100} - \quad (4 \times 64 \text{ bits}) \equiv 4w \text{ bits. dictionary}$$

a a c a

\rightarrow abf ggadf had(a bb gg ah aca) hh bcdeacd : dfgha abfgga

$n=2$ $n=8$
 $\lambda=5$ $\lambda=12, \lambda=13$
 $n=5 \leftarrow \max$ $n=3$ $n=5$
 $\lambda=31$

$\lambda=38$ Run length
 \downarrow \downarrow
 $n=6, \lambda=38 \equiv \frac{000000100110}{\lambda} \quad \frac{000110}{n} \equiv abfgga$

$$n=2, \lambda=5 \equiv \frac{000101}{\lambda} \quad \frac{0010}{n} \equiv df$$

$$n=3, \lambda=31 \equiv$$

<u>000000</u>	<u>100110</u>	<u>000110</u>	<u>000101</u>	<u>0010</u>	<u>0000011111</u>	<u>0011</u>	<u>1111</u>
							<u>h</u>
<u>00001100</u>	<u>00001000</u>						

- * If there is no match, we go with defined trivial coding
- * we can differentiate btw trivial coding and run length - w matching by first bit.

First bit : 0 - match in w

1 - trivial coding

— w — $\xrightarrow{\text{sequence}}$

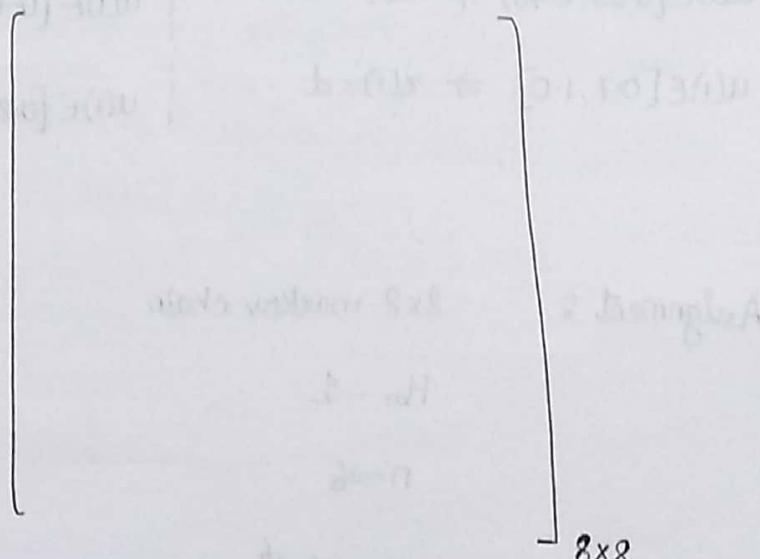
* --- dfghaabfgga df had hh bcdeacd dfghaabfgga

$\downarrow n=5$

$n=13$ (there is no upper bound)

000101 00001011
l n

→ Markov Chain:



→ ergodic.

Q:

$$a \begin{bmatrix} 0.25 & 0.30 & 0.15 & 0.30 \\ 0.50 & 0.20 & 0.10 & 0.20 \\ 0.10 & 0.40 & 0.40 & 0.10 \\ 0.60 & 0.10 & 0.15 & 0.15 \end{bmatrix}$$

$$u = \text{rand}[0,1]$$

$u \in [0, 0.25) \Rightarrow \text{choose } x(0) = a$

$u \in [0.25, 0.5) \Rightarrow \text{choose } x(0) = b$

$u \in [0.5, 0.75) \Rightarrow \text{choose } x(0) = c$

$u \in [0.75, 1.0] \Rightarrow \text{choose } x(0) = d$

if $x(0) = a$

$u(1) = \text{rand}[0,1]$

$u(1) \in [0, 0.25) \Rightarrow x(1) = a$

$u(1) \in [0.25, 0.55) \Rightarrow x(1) = b$

$u(1) \in [0.55, 0.70) \Rightarrow x(1) = c$

$u(1) \in [0.7, 1.0] \Rightarrow x(1) = d$

if $x(0) = b$

$u(1) = \text{rand}[0,1]$

$u(1) \in [0, 0.5) \Rightarrow x(1) = a$

$u(1) \in [0.5, 0.7] \Rightarrow x(1) = b$

$u(1) \in [0.7, 0.8) \Rightarrow x(1) = c$

$u(1) \in [0.8, 1.0] \Rightarrow x(1) = d$

Assignment : 8x8 markov chain

$$H_{\infty} = 1$$

$$n \sim 6$$

$$\omega = 6 \cdot 2^6 \sim 1000$$

$$L = 1000000$$

- ① Markov chain coding
 ② Z-L encoding
 ③ Z-L decoding

} 3 parts of code

→ Bernoulli Random Variable : $P(H) = p$, $P(T) = q$, n trials.

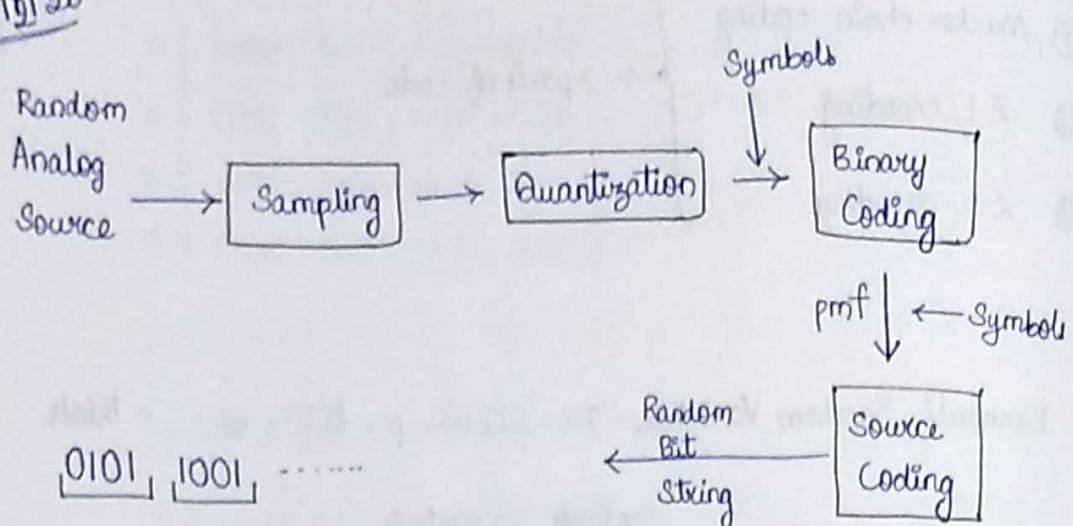
typical event : $\underbrace{\text{np heads}}_{(p)^{np}}, \underbrace{\text{nq tails}}_{(q)^{nq}}$

$$= 2^{\log_2(p^{np} \cdot q^{nq})}$$

$$= 2^{n[p \log_2 p + q \log_2 q]}$$

$$= 2^{-nH}$$

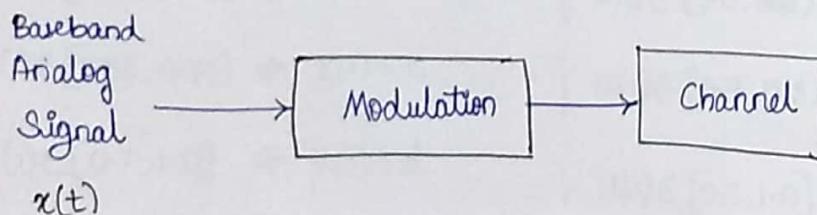
28/9/20



Channel Bandpass

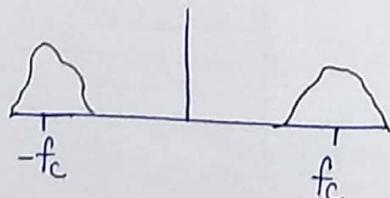
Rx Receiver

* Analog Communication :



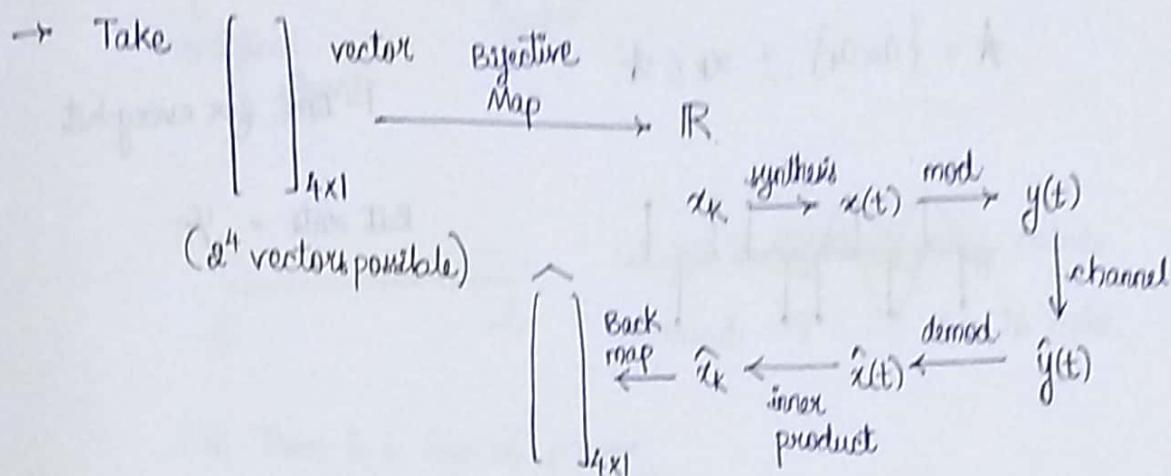
$$y(t) = R \left[x(t) \cdot e^{j2\pi f_c t} \right]$$

* Channel : Bandpass



- { • Attenuation
- Multipath
- Group delay - Phase delay } \rightarrow Distortion equalization
- Noise.

$$\rightarrow x(t) = \sum_{k \in \mathbb{Z}} x_k \phi_k(t), \quad x_k = \langle x(t), \phi_k(t) \rangle$$



→ Mostly random analog source signal is redundant, so modulating completely is not necessary. We just want to modulate the changes occurred i.e., modulating only INFORMATION. Now there is no redundancy.

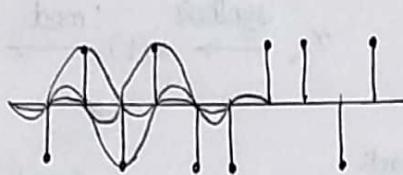
→ Total 4 conversions : $\overset{\text{I/P}}{\text{A/D}} \rightarrow \text{D/A} \rightarrow \overset{\text{O/P}}{\text{A/D}} \rightarrow \text{D/A} \rightarrow \overset{\text{O/P}}{\text{D/P}}$
(in digital communication)

→ Digital modulation : mapping ; synthesis, modulation.

→ 0101001101 1 bit mapping $0 \rightarrow -5$ $a_0 = -5$
 $1 \rightarrow +5$ $a_1 = +5$

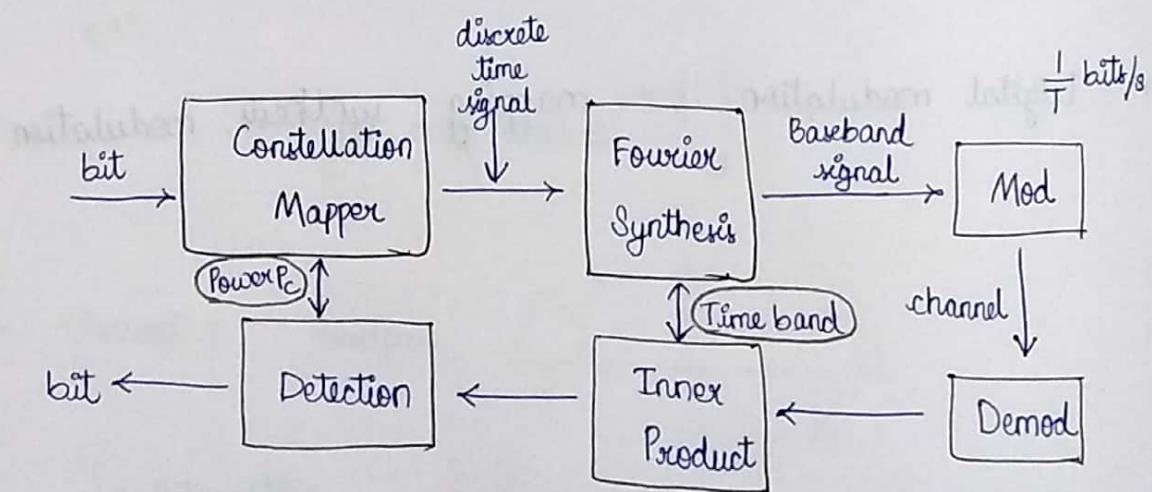
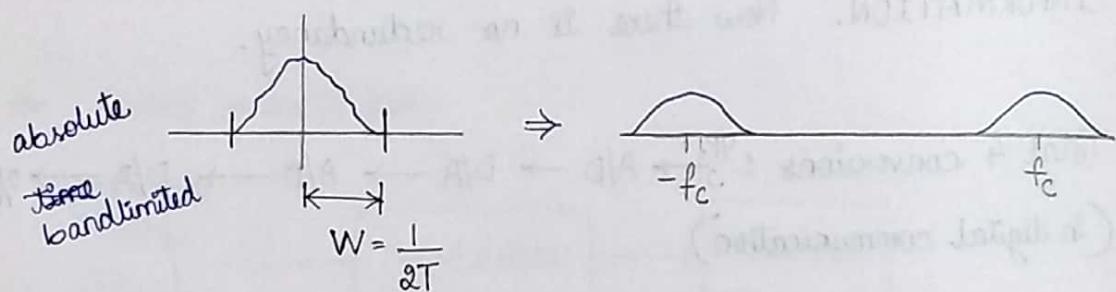
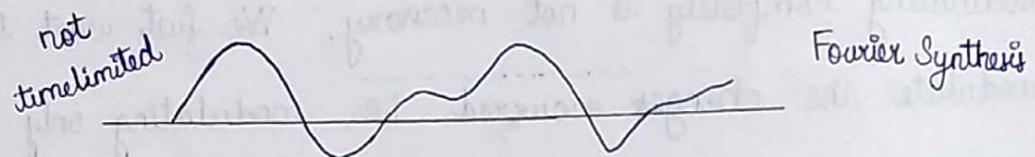
$$A = \{a_0, a_1\} ; x_k \in A$$

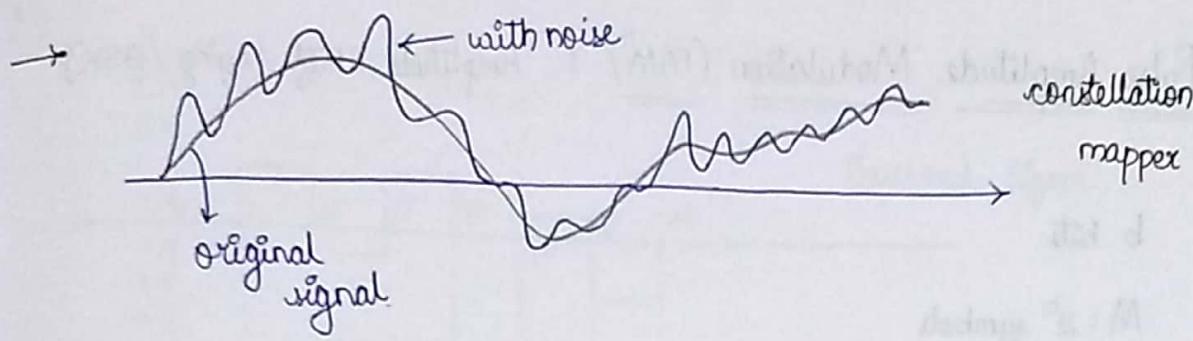
T-time for every bit



Bit rate = $\frac{1}{T}$

$$x(t) = \sum_{k \in \mathbb{Z}} x_k \phi_k(t) = \sum x_k \cdot \text{sinc}\left(\frac{t}{T} - k\right)$$

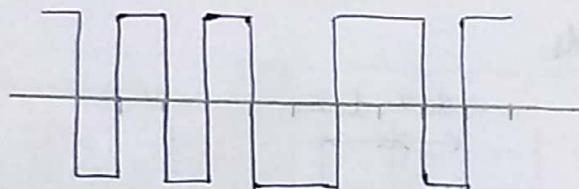




we get a distribution of values around $+5 \& -5$ due to noise.

- * There is a loss of power.

→ choosing $\phi_k(t)$ as rectangular pulse



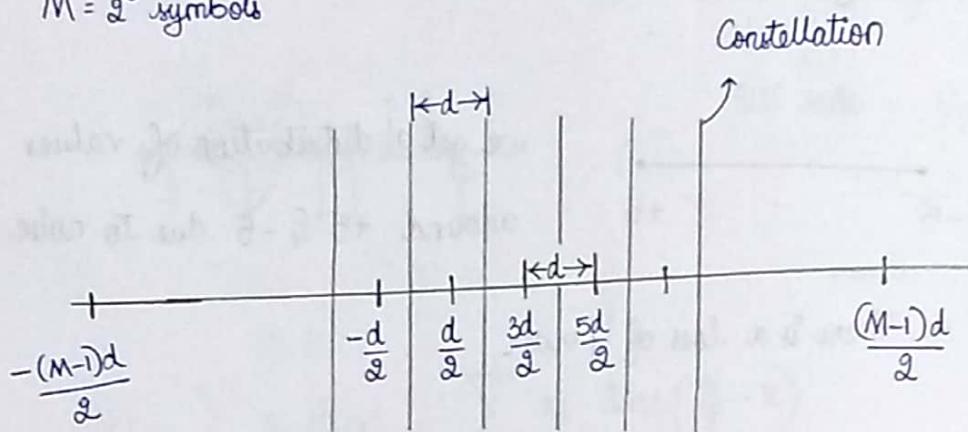
timelimited
but infinite bandwidth

- * Sinc & Rect are two extremes of mapping
- * So we take approximately timelimited & bandlimited

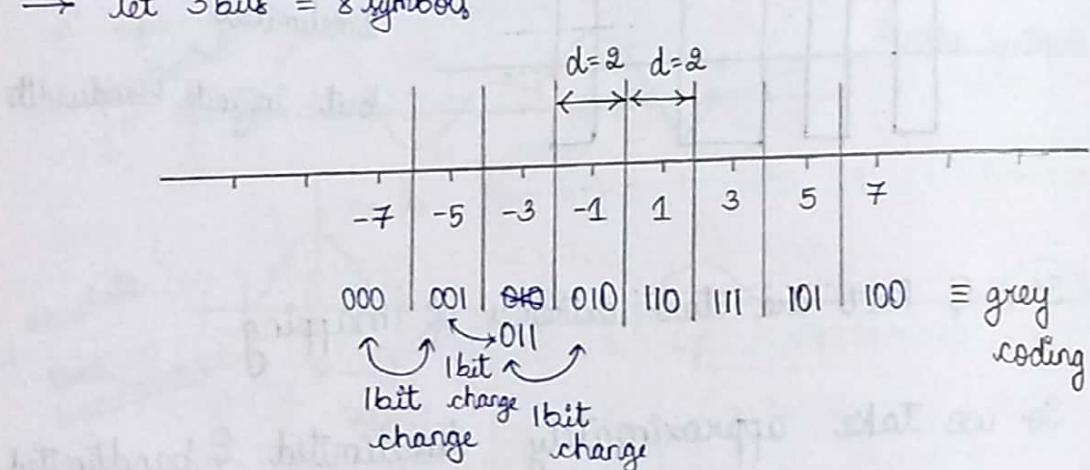
Pulse Amplitude Modulation (PAM) : Amplitude Shift Keying (ASK)

b bits

$$M = 2^b \text{ symbols}$$



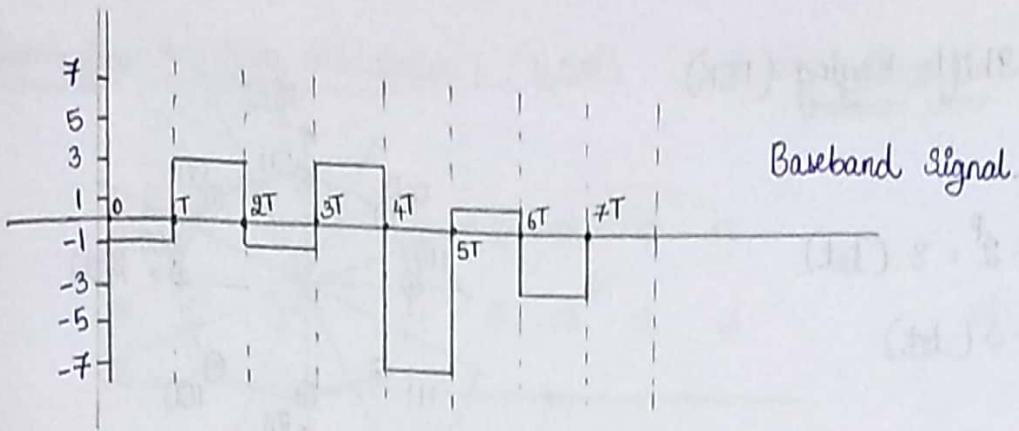
→ Let 3 bits \equiv 8 symbols



$$\begin{matrix} 010 & 111 & 010 & 111 & 000 & 110 & 011 \\ -1 & 3 & -1 & 3 & -7 & 1 & -3 \end{matrix}$$

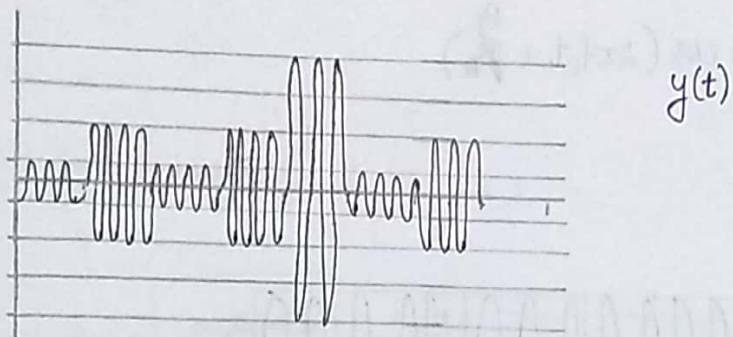
$$\phi_k(t) = u(t-kT) - u(t-(k+1)T)$$

$$x(t) = \sum x_k \phi_k(t)$$



$$\rightarrow y(t) = \Re \left[x(t) \cdot e^{j2\pi f_c t} \right]$$

$$= \sum_k x_k \phi_k(t) \cos(2\pi f_c t)$$

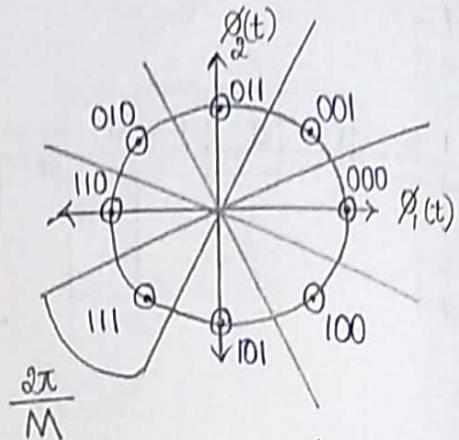


→ Read about Power consumption from book

Phase Shift Keying (PSK)

$$M = 2^P = 8 \text{ (let)}$$

$$P = 3 \text{ (let)}$$

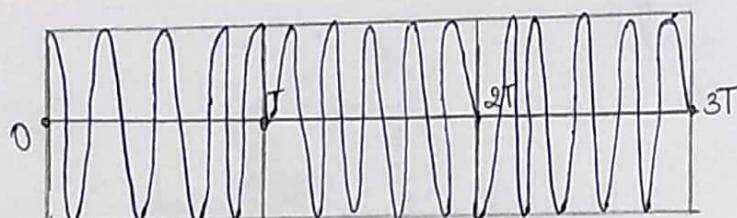


$$\theta_k = \frac{2\pi}{M} k$$

$$y(t) = R \left[A e^{j\theta_k} e^{j2\pi f_c t} \right]$$

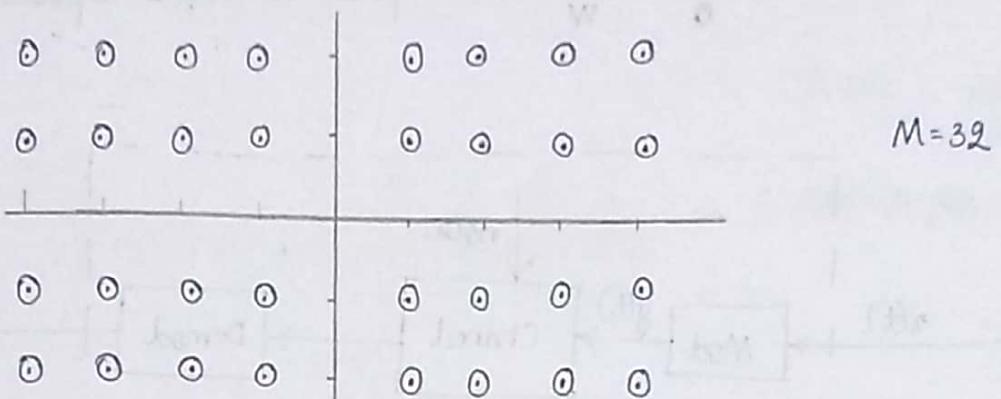
$$= A \cos \theta_k \cos(2\pi f_c t) - A \sin \theta_k \sin(2\pi f_c t)$$

$$= A \cos(2\pi f_c t + \theta_k)$$



→ Read about Power Consumption

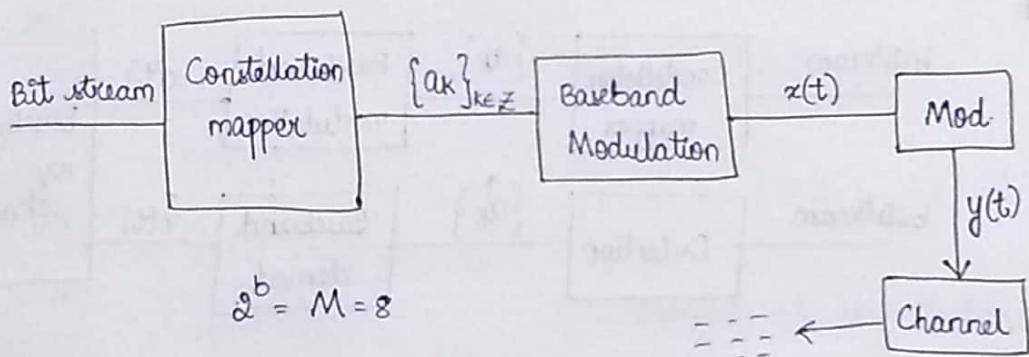
Quadrature Amplitude Modulation : (QAM) 2 Dimensional



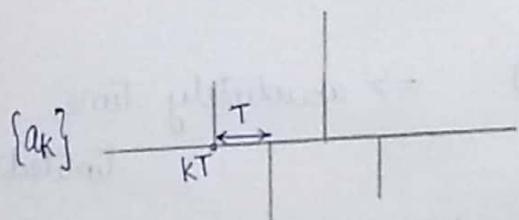
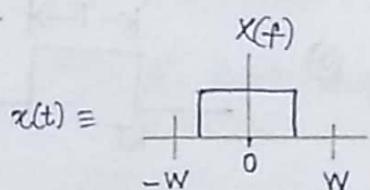
$$y(t) = A_{ki} \cos(2\pi f_c t) + A_{kq} \sin(2\pi f_c t)$$

4/3/20

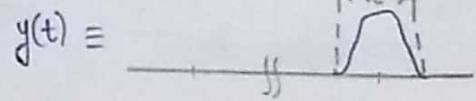
PAM : (ASK)



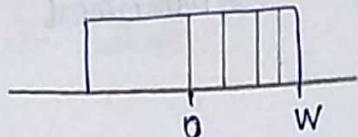
$$2^b = M = 8$$



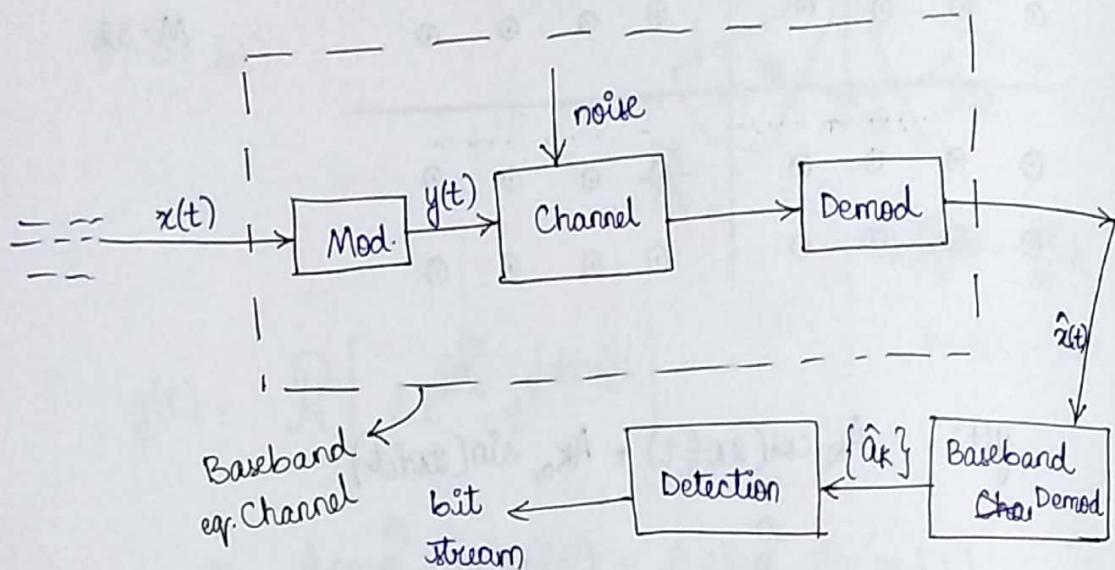
{ AM, DSB-SC,
SSB, PM, FM }



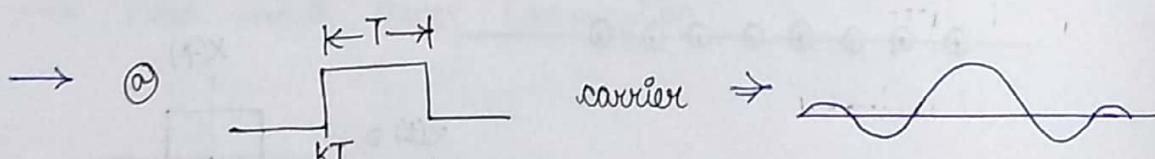
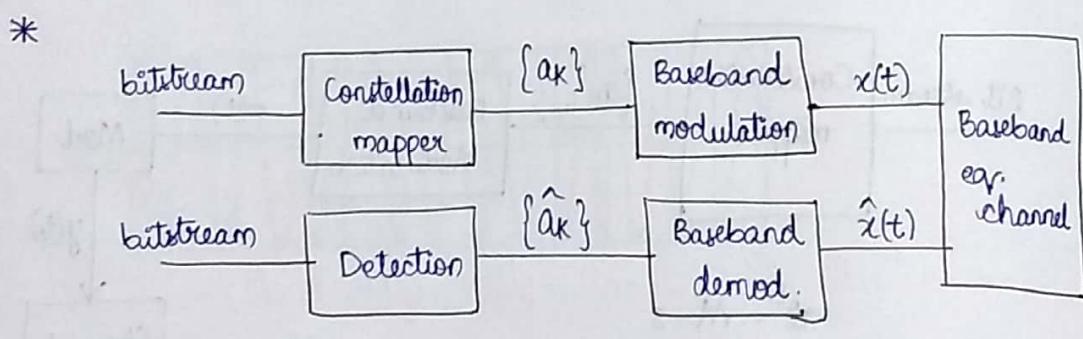
FSK :



→ Splitting the band and each part is dealt separately.



→ These mappings are not linear.

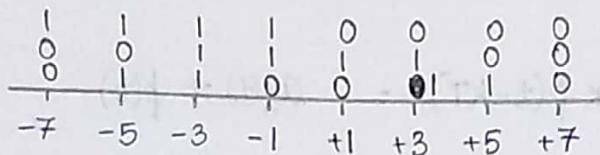


$$x(t) = \sum_{k \in \mathbb{Z}} a_k u(t - kT) \Rightarrow \text{absolutely time limited}$$

⇒ but not absolutely band limited

- Sampling at any instant in a time window $\equiv T$ always gives a defined value back sequence.

*

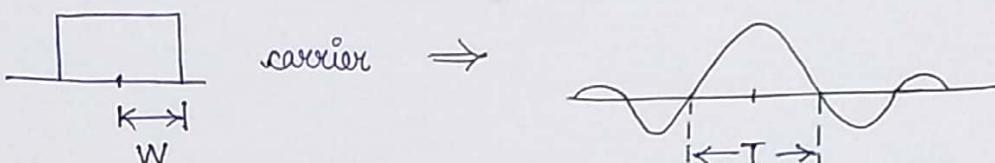


(V)

$$T = 1 \text{ ms} \equiv \text{clk is } 1 \text{ kHz}$$

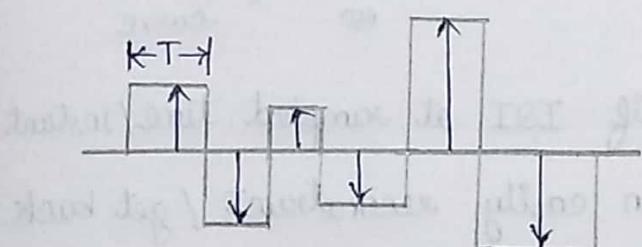
3 kilobits per sec

→ ⑥



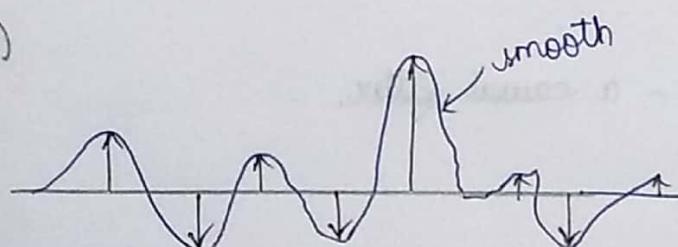
- Here we need to sample at exact instant kT , then only we get back sequence.

(a)

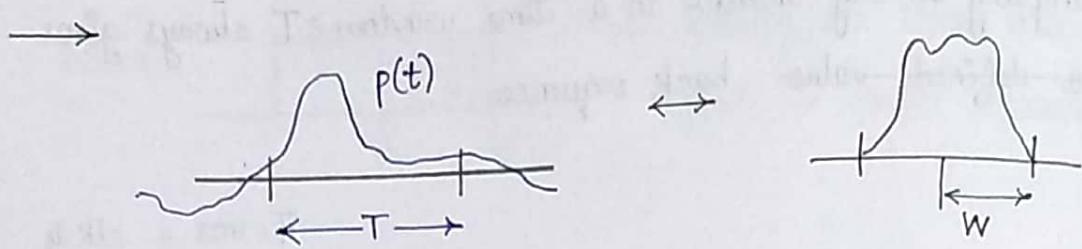


Sample at any instant in time window T , we can get back sequence

(b)



Sample at exact instant, we can get back sequence



$$x(t) = \sum_{k \in \mathbb{Z}} a_k p(t - kT) = a_s(t) * p(t)$$

where $a_s(t) = \sum_{k \in \mathbb{Z}} a_k \delta(t - kT)$

$x(t) \equiv$ filtering of $a_s(t)$ with $P(f)$.

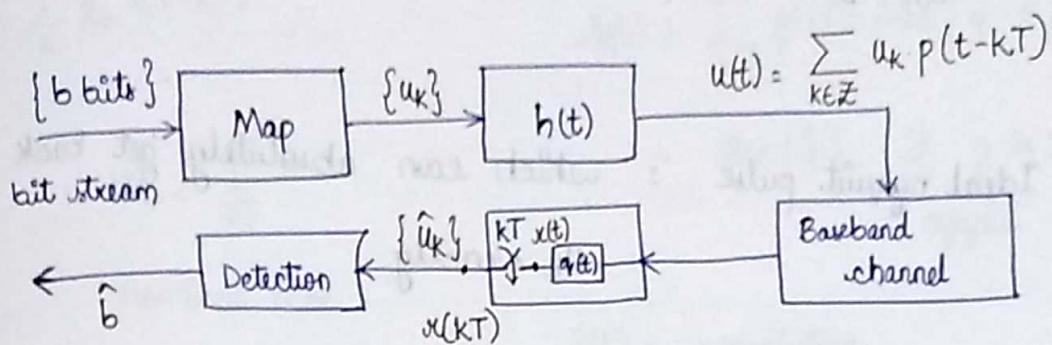
ISI : Inter Symbol Interference

- at an instant, the signal value is not completely due to that symbol, it is sum of effects of symbols in that time window,
- This is also possible if $p(t) \equiv \text{sinc}(t)$.
an ideal nyquist
PF curve
- Now while sampling, if ISI at sampled time/instant is "zero" then we can easily reconstruct / get back the sequence.

Sinc - a causal filter.

11/3/20

Nyquist Criterion / Matched Filter :



Linear Baseband Decoding.

$$r(t) = q_r(t) * u(t) = \int_{-\infty}^{\infty} u(\tau) \cdot q_r(t-\tau) d\tau$$

$$= \int_{-\infty}^{\infty} \left[\sum_{k \in \mathbb{Z}} u_k p(\tau - kT) \right] \cdot q_r(t-\tau) d\tau$$

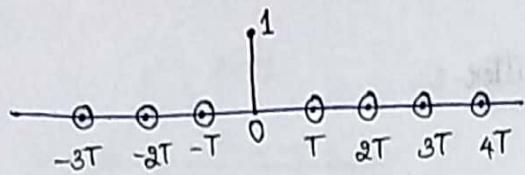
$$= \sum_{k \in \mathbb{Z}} u_k \int_{-\infty}^{\infty} p(\tau - kT) \cdot q_r(t-\tau) d\tau$$

provided $g = p * q_r = \int_{-\infty}^{\infty} p(\tau) q_r(t-\tau) d\tau \in L_2$

$$= \sum_{k \in \mathbb{Z}} u_k g(t - kT)$$

$$x(KT) = \sum_{m \in \mathbb{Z}} u_k g((K-m)T) = u_k \text{ iff } g(KT) = \delta_k$$

Restriction is on
Function values at nT ,

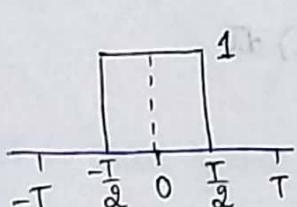


≡ Ideal nyquist pulse : which can absolutely get back u_k exactly.

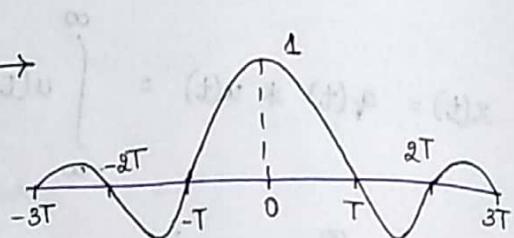
→ in this case we say band matched.

→ we can recover the signal without ISI

→



Ideal nyquist
curve



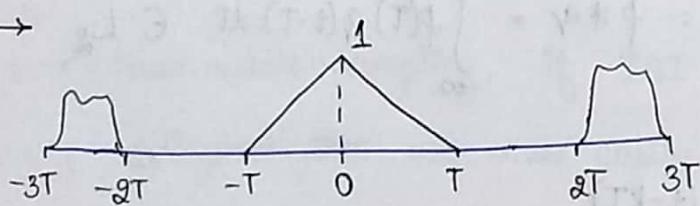
Ideal nyquist
curve

→ infinite B/W

→ synchronization issue

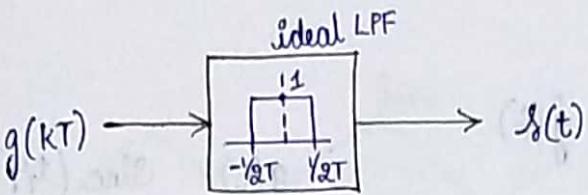
→ dies down very slowly.

→



Ideal nyquist curve

→ can have differentiability issues



$$= \sum_k g_k \delta(t - kT)$$

$$\delta(t) = \sum_k g(kT) \cdot \text{sinc}\left(\frac{t}{T} - k\right)$$

$= \text{sinc}\left(\frac{t}{T}\right)$ if g is ideal
Nyquist.

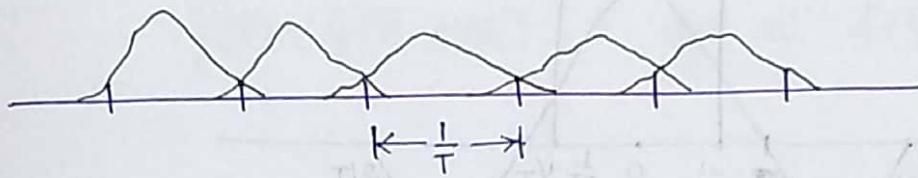
$$W_b = \frac{1}{2T}$$

nominal B/W

$\delta(t) = \text{convolution}$

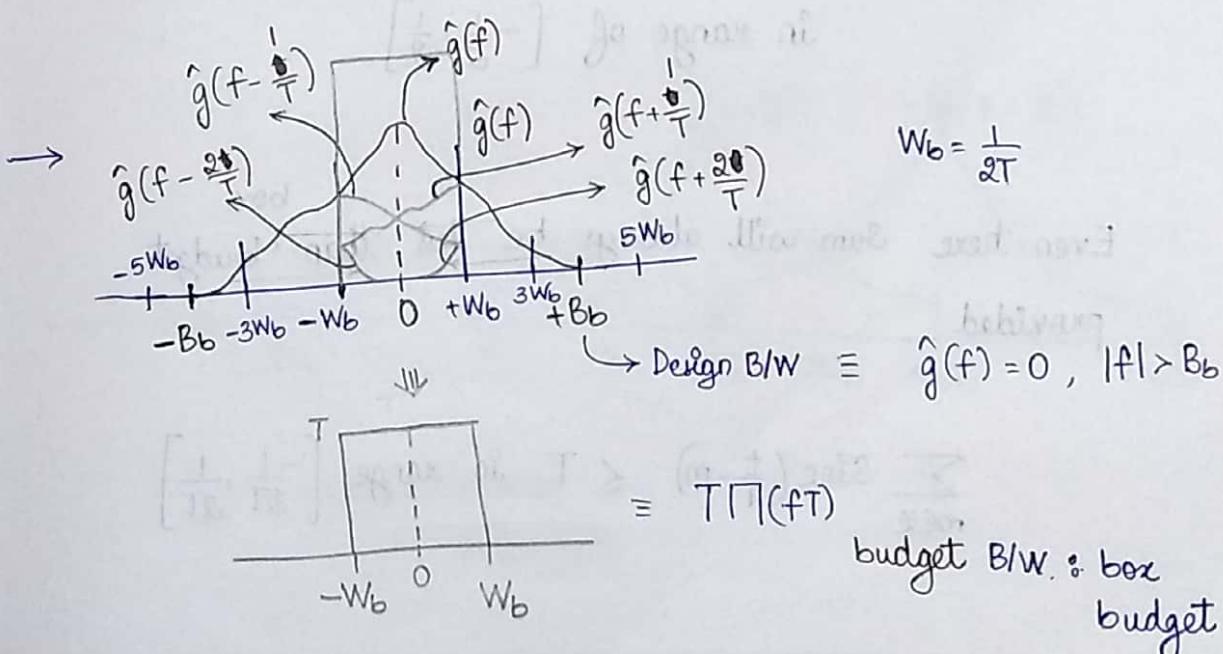
$$\sum g(kT) * \text{sinc}\left(\frac{t}{T}\right)$$

$$\left[\sum_m \hat{g}\left(f + \frac{m}{T}\right) \right] \Pi(fT) = T \Pi(fT)$$

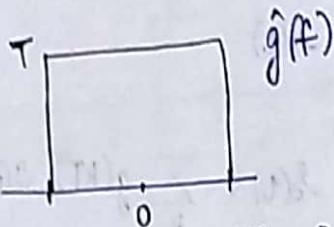


* L_2 convergence : $\int_{-\infty}^{\infty} \left[\sum_m \hat{g}\left(f + \frac{m}{T}\right) \cdot \Pi(fT) - T \Pi(fT) \right]^2 df = 0$

→ aliasing is happening.



→



$$g(t) = \text{sinc}(t/T)$$

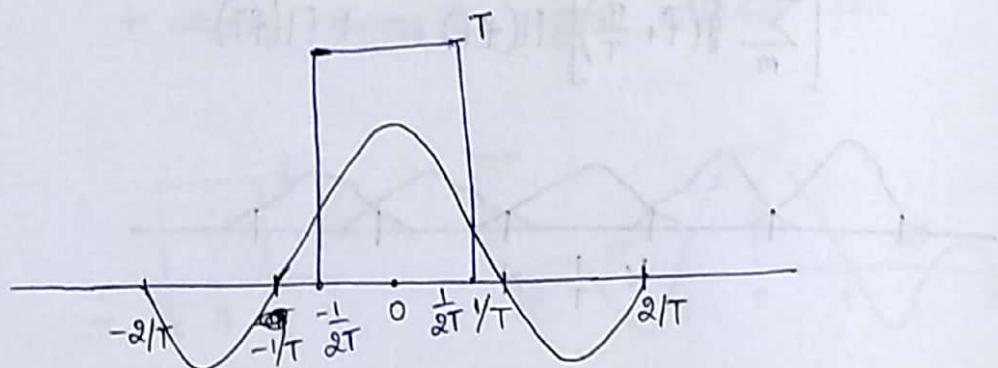
→ very much Nyquist

$$W_b = B_b = \frac{1}{2T}$$

nominal B/W = design B/W.

Entire budget B/W is used.
box budget

→



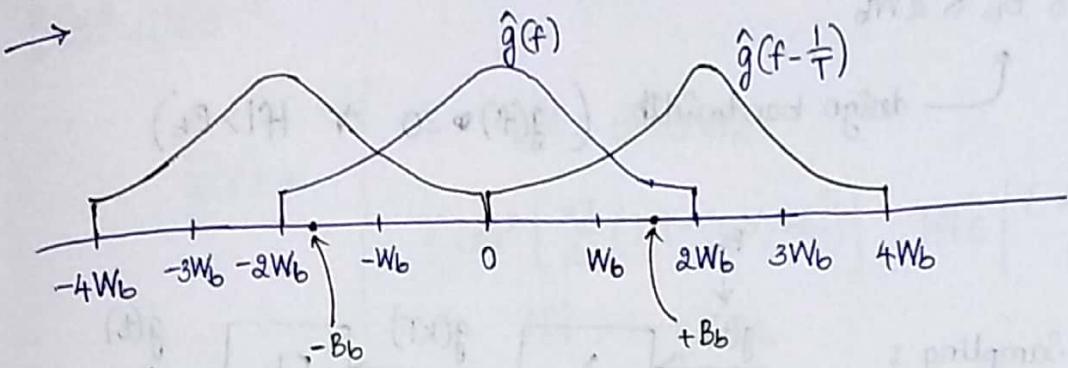
check whether $\text{sinc}(t) + \text{sinc}(t-1) + \text{sinc}(t-2) + \dots$

$$= \sum_{n=-5}^{5} \text{sinc}(t+n)$$

in range of $[-\frac{1}{2}, \frac{1}{2}]$

Even here sum will always be less than ^{box} budget provided

$$\sum_{m \in \mathbb{Z}} \text{sinc}\left(\frac{t-m}{T}\right) \leq T \text{ in range } \left[-\frac{1}{2T}, \frac{1}{2T}\right]$$



$$\sum_{m=0,-1} \hat{g}(f + mW_b) \Pi(fT) = T \Pi(fT) \quad W_b < B_b \leq 2W_b$$

maximum number of overlaps = 2.

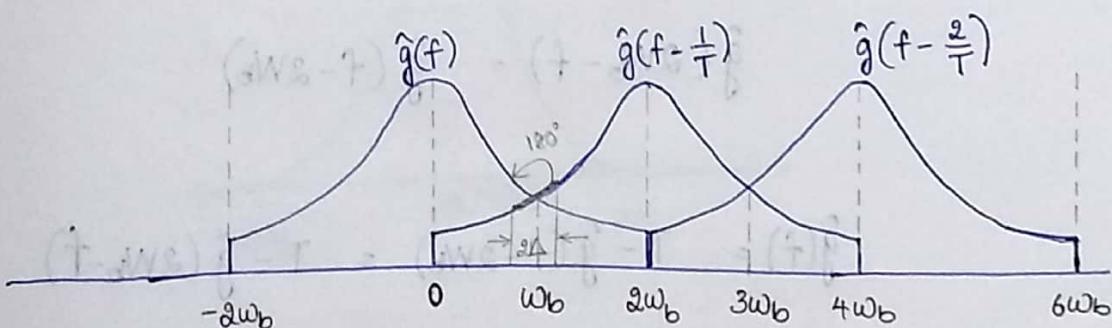
$$\hat{g}(f) + \hat{g}(f - 2W_b) = T \text{ for all } f \in [0, W_b]$$

12/3/20

Nyquist criterion:

$$\text{lim} \sum_{m \in \mathbb{Z}} \hat{g}\left(f + \frac{m}{T}\right) \cdot \Pi(fT) = T \Pi(fT)$$

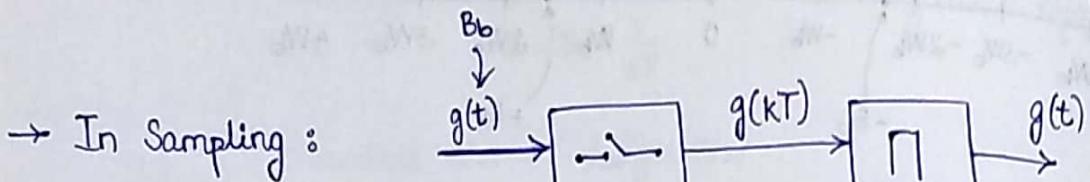
limit in mean square (L_2)



$$\frac{1}{2T} = W_b \leftarrow \text{nominal bandwidth}$$

$$W_b \leq B_b \leq 2W_b$$

design bandwidth ($|g(f)| = 0 \text{ if } |f| > B_b$)



$$W_b = \frac{1}{2T} \geq B_b \Rightarrow \text{no aliasing}$$

\rightarrow But here we are doing exact opposite ; that is we are using aliasing.

$$W_b \leq B_b \leq 2W_b$$

$$\therefore \hat{g}(f) + \hat{g}(f - 2W_b) = T \quad \forall f \in [0, W_b]$$

we know

we take $g(t)$ - real & even $\Rightarrow \hat{g}(f)$ - real & even

$$\hat{g}(2W_b - f) = \hat{g}(f - 2W_b)$$

$$\hat{g}(f) = T - \hat{g}(f - 2W_b) = T - \hat{g}(2W_b - f)$$

$$\text{let } W_b - f = \Delta$$

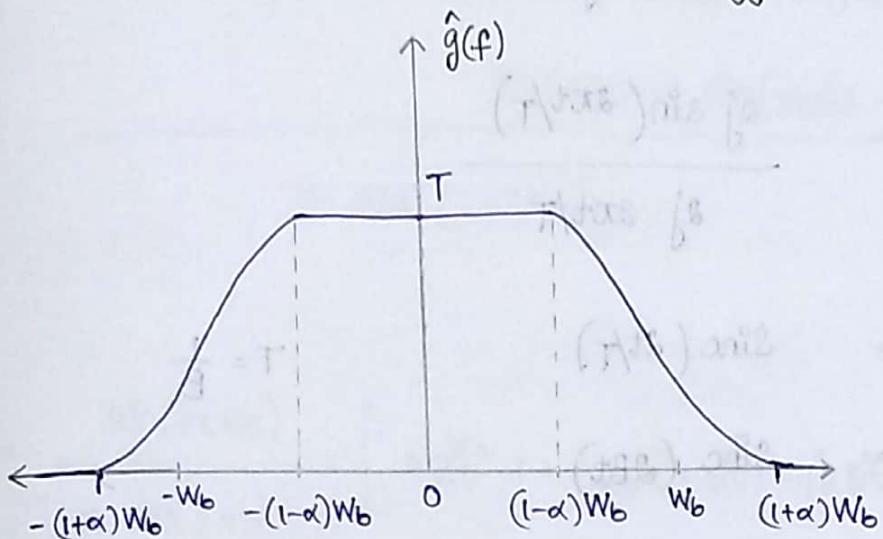
$$\boxed{\hat{g}(W_b - \Delta) = T - \hat{g}(W_b + \Delta)}$$

Rotational symmetry

Raised Cosine :

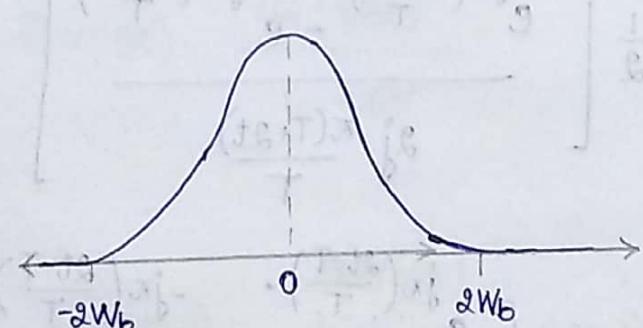
$$\hat{g}(f) = \begin{cases} T, & |f| \leq (1-\alpha)W_b \\ T \cos^2 \left[\frac{\pi T}{2\alpha} (|f| - (1-\alpha)W_b) \right], & |f| \in \left[(1-\alpha)W_b, (1+\alpha)W_b \right] \\ 0, & |f| \geq (1+\alpha)W_b \end{cases}$$

↑ roll off



* consider $\alpha = 1\%$ $\hat{g}(f) = T \cos^2 \left(\frac{\pi T}{2\alpha} f \right)$

$$= \frac{T}{2} (1 + \cos(\pi T f))$$



$$g(t) = \int_{-\infty}^{\infty} \hat{g}(f) e^{j 2\pi f t} df$$

$$= \frac{I}{2} \int_{-2W_b}^{2W_b} [1 + \cos(\pi f T)] e^{j 2\pi f t} \cdot df$$

$$1^{\text{st}} \text{ term} = \frac{I}{2} \cdot \frac{e^{j 2\pi t/T} - e^{-j 2\pi t/T}}{j 2\pi t}$$

$$= \frac{2j \sin(\pi t/T)}{2j \cdot \pi t/T}$$

$$= \text{sinc}(\pi t/T) \quad T = \frac{1}{B}$$

$$= \text{sinc}(2Bt)$$

$$2^{\text{nd}} \text{ term} = \frac{I}{2} \int_{-1/T}^{1/T} \left(\frac{e^{j\pi f T} + e^{-j\pi f T}}{2} \right) e^{j 2\pi f t} \cdot df$$

$$= \frac{1}{2} \left[\frac{e^{j\pi(T+2t)/T} - e^{-j\pi(T+2t)/T}}{2j \pi(T+2t)/T} \right]$$

$$+ \frac{1}{2} \left[\frac{e^{j\pi(2t-T)/T} - e^{-j\pi(2t-T)/T}}{2j \pi(2t-T)/T} \right]$$

$$= \frac{1}{2} \text{sinc}(2Bt+1) + \frac{1}{2} \text{sinc}(2Bt-1)$$

$$\therefore g(t) = \text{sinc}(2Bt) + \frac{1}{2} \text{sinc}(2Bt+1) + \frac{1}{2} \text{sinc}(2Bt-1)$$

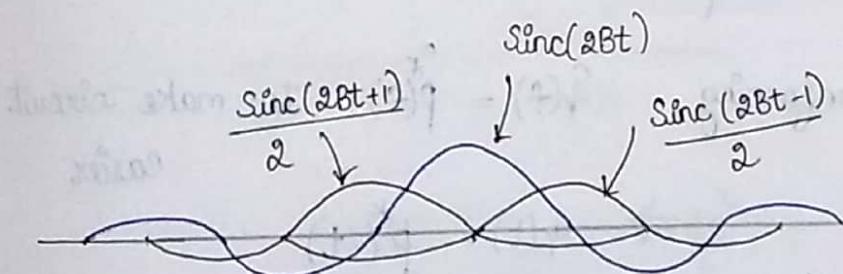
$$2W_b = B = \frac{1}{T}$$

$$g(t) = \frac{(4Bt-1) \sin(2\pi Bt) + Bt(2Bt-1) \sin(2\pi Bt+\pi) - Bt(2Bt+1) \sin(2\pi Bt-\pi)}{\pi 2Bt (4Bt-1)}$$

$$= \frac{\sin(2\pi Bt)}{\pi 2Bt (4Bt-1)} \left[4Bt-1 - (2Bt+Bt) - (2Bt-Bt) \right]$$

$$= \frac{\sin(2\pi Bt)}{\pi 2Bt (1-4Bt^2)} = \frac{2 \sin(\pi Bt) \cos(\pi Bt)}{\pi \pi Bt (1-4Bt^2)}$$

$$= \text{sinc}(Bt) \cdot \frac{\cos(\pi Bt)}{1-4Bt^2}$$



for general α :
$$g(t) = \frac{\sin(\alpha Bt) \cos(\pi \alpha Bt)}{1 - 4B^2 \alpha^2 t^2}$$

$g(t)$ is strictly bandlimited but not timelimited.

We need to show :

$$u(t) = \sum u_k p(t-kT) \equiv u(t) * q(t) \text{ } \underline{kT}$$

$$u_k = \langle u(t), p(t-kT) \rangle \text{ equivalent to}$$

[multiplying and
integration]

high analog cost

[filtering and
sampling]

low analog cost

Let

$$\rightarrow \hat{g}(f) = \hat{p}(f) \cdot \hat{q}(f)$$

$\hat{p}(f) \leftarrow \text{Complex}$

we are enforcing $\hat{q}(f) = \hat{p}^*(f)$ to make circuit easier

$$\Rightarrow q(t) = p^*(-t)$$

$$\hat{g}(f) = |\hat{p}(f)|^2$$

$$g(kT) = \int_{-\infty}^{\infty} p(t) \cdot g(kT-t) \cdot dt$$

$$= \int_{-\infty}^{\infty} p(t) \cdot p^*(t-kT) \cdot dt$$

$$= \langle p(t), p(t-kT) \rangle = \delta_k$$

↓ Nyquist condition

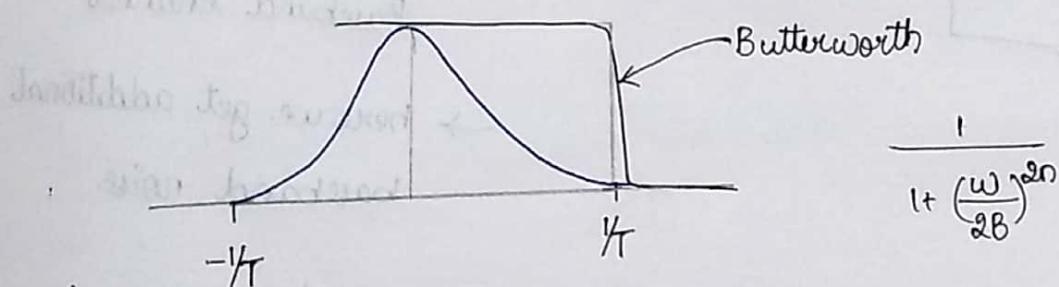
∴ $\{p(t-kT)\}$ set is orthonormal basis.

Nyquist condition \Rightarrow orthonormal basis.

13/3/20

$$\hat{g}(f) = \frac{T}{2} \left(1 + \cos\left(\frac{\pi\omega}{2}\right) \right) = \frac{T}{2} \left(1 + \cos(\pi f T) \right)$$

realization can be done by FIR / IIR



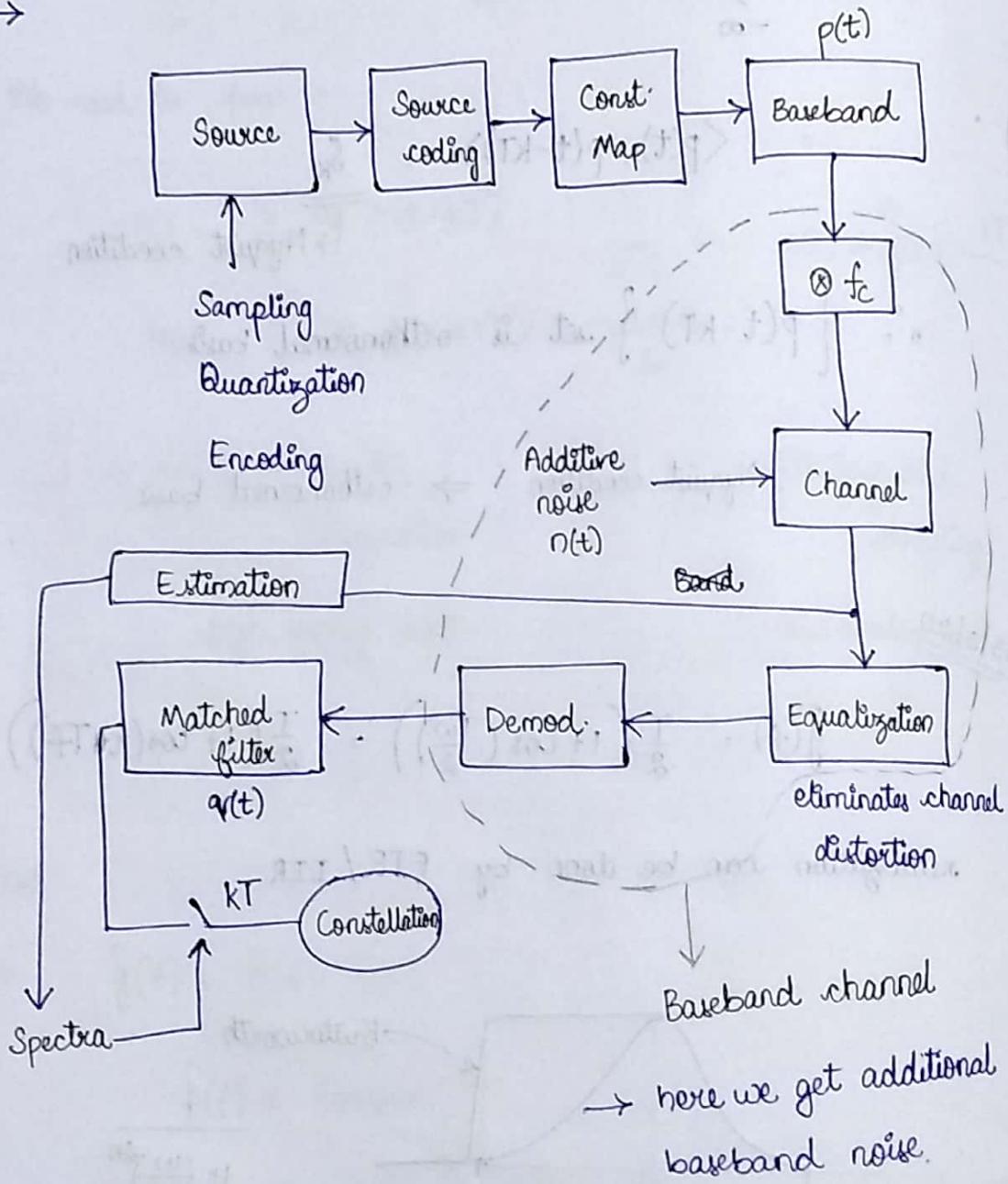
$$\frac{1}{1 + \left(\frac{\omega}{2B}\right)^{2n}}$$

$$\frac{\omega_0}{B} = \frac{\pi}{T}$$

$$\hat{g}(f) = \frac{1}{1 + \left(\frac{\omega}{\omega_R}\right)^{2n}} \int_{-\infty}^{\infty} 1 + 1 - \frac{\pi\omega}{B}$$

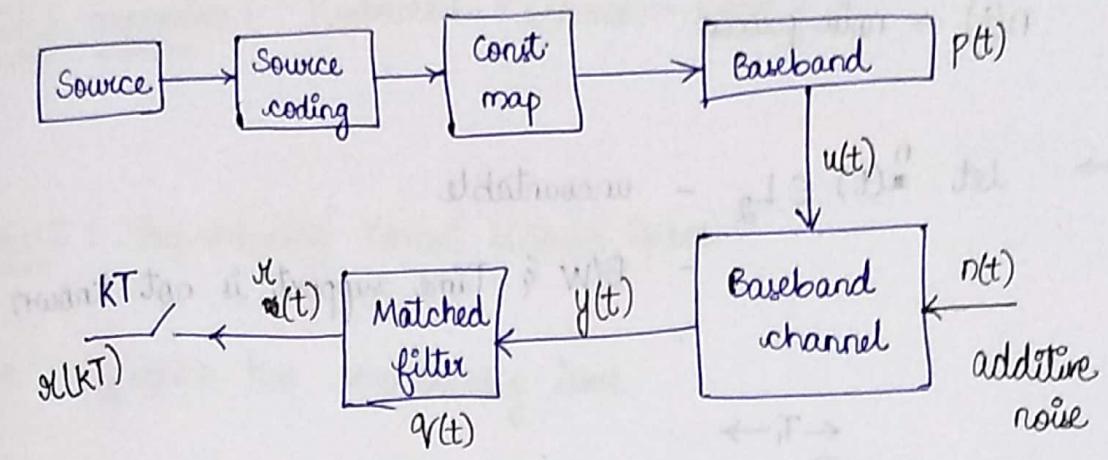
READ in PROAKIS :

- Modulation :
- (i) Map : energy & prob. of error.
 - (ii) Continuous phase modulation



→ Until now, we had ^{only} quantization noise; now we have another noise \equiv Baseband noise.

Nyquist $\rightarrow g(t) = p(t) * q_r(t)$ (without $n(t)$)



$$u(t) = \sum_k u_k \cdot p(t - kT)$$

$$y(t) = u(t) + n(t) = \sum_k u_k \cdot p(t - kT) + n(t)$$

$$x(t) = y(t) * g(t) = \sum_k u_k g(t - kT) + \int n(\tau) g(t - \tau) d\tau$$

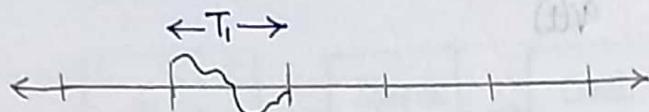
$$\mathcal{H}(kT) = U_k + \eta_k$$

where $D_K = \int n(\tau) \cdot q_r(KT - \tau) \cdot d\tau$

$n(t) \rightarrow$ noise process.

→ Let $\phi(t) \in L_2$ - uncountable

- B/W & Time support is not known.



$n(t), t \in T_1$

$$n(t) = \sum_{k \in \mathbb{Z}} n_k \phi_k(t), t \in T_1$$

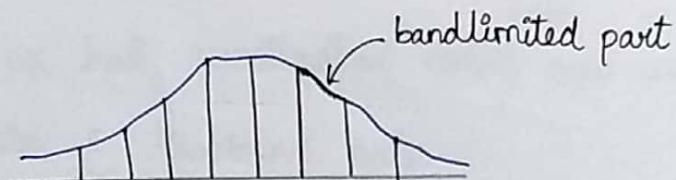
$$n_k = \langle n(t), \phi_k(t) \rangle = \int_T \dots$$

random variable

in each truncated window, we apply F.S in that window. We get a set of countable coefficients in each window.

This way uncountable \Rightarrow countable

→ Similarly



each such part has an inverse in time domain.

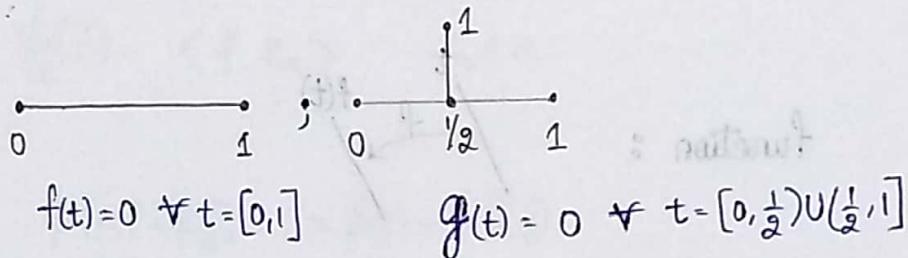
[KKL expansion : Kosthambi - Karhunen - Loeve.]

RKHS : Reproducing Kernel Hilbert Space.

→ L_2 space has smoothening issue.

$$\|f-g\|_2 = 0$$

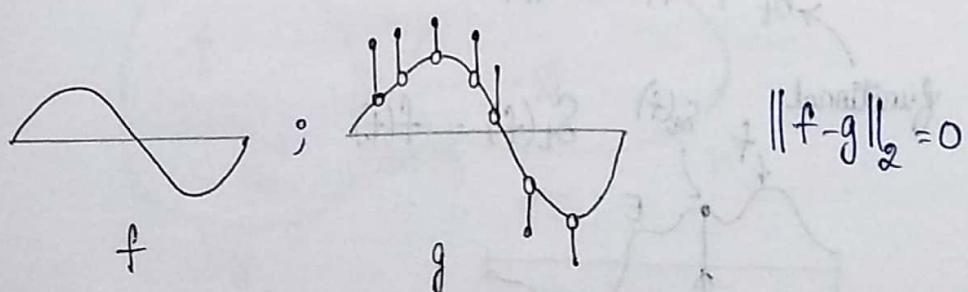
Let



$\|f-g\|_2 = 0$. : L_2 distance is zero but both functions are not same.

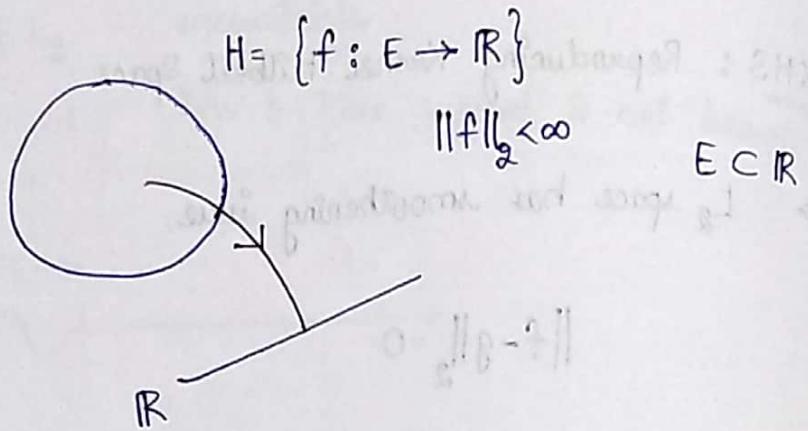
∴ L_2 convergence is not point wise convergence; in case of spikes involved

also

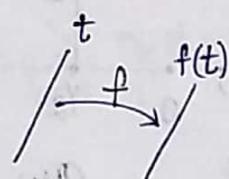


→ So we need a smooth Hilbert space.

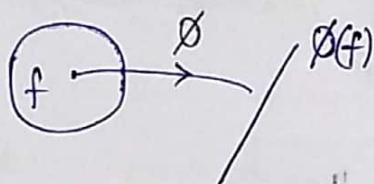
Functional :



function :

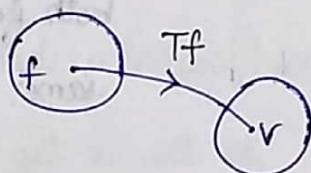


functional :



∞ no. of functionals
can be defined
on H.

operator :

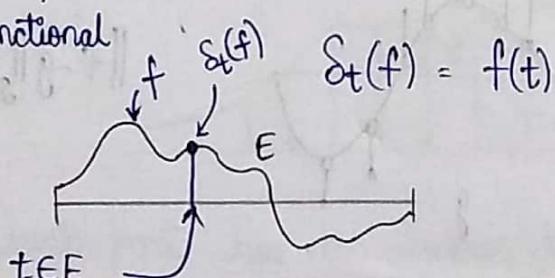


geometrical view of linear functionals

Dirac Functional :

$s_t : H \rightarrow \mathbb{R}$

functional



$$s_t(f) = f(t)$$

$$\text{eg: } \int_E \delta(\tau-t) f(\tau) d\tau = f(t)$$

$$\delta(t) * f = f(t)$$

\uparrow this particular dirac delta functional is not continuous

Riesz representation theorem:

Any linear functional (continuous) ϕ on f can be written as inner product.

$$\phi_x(f) = \langle f, K_x \rangle \quad K_x \in H.$$

$$\text{linear: } \phi(\alpha f + \beta g) = \alpha \phi(f) + \beta \phi(g)$$

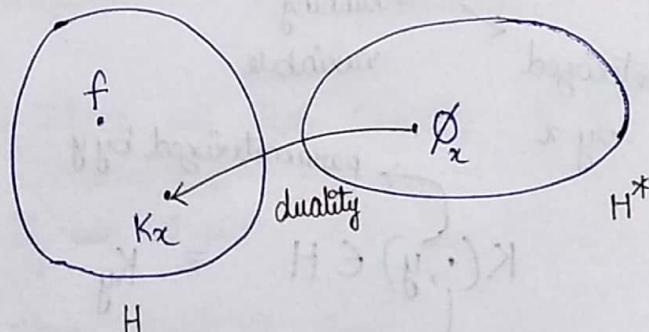
$$\text{continuity: } \|f - g\|_2 < \epsilon \Rightarrow |\phi(f) - \phi(g)| < \delta$$

$\overset{\text{continuous}}{\text{if we use dirac functional and shift it along } (-\infty, \infty)}$

L_2 convergence will imply point wise convergence.

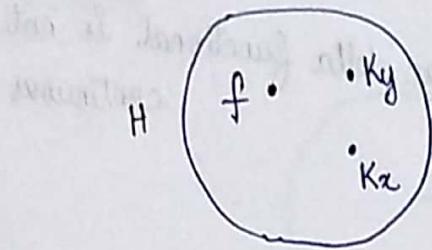
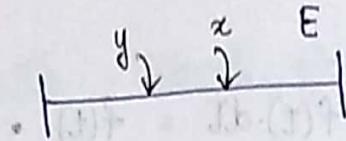
* for smooth functions : L_2 convergence \Rightarrow Point wise convergence.

*





$$E \in \mathbb{R}$$

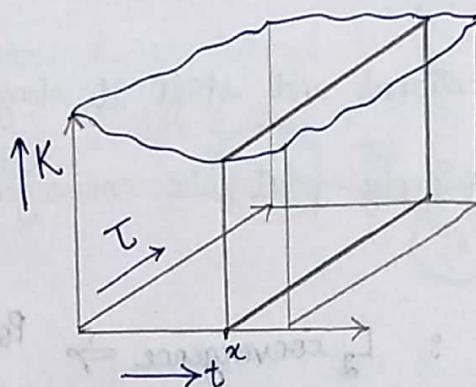


$$\phi_y(K_x) = \langle K_x, K_y \rangle$$

$= K(x, y) \rightarrow \text{kernel}$

$$K: E \times E \rightarrow \mathbb{R} \quad (\text{function})$$

$K_x \rightarrow$ fix x in K



$$K(t, T): E \times E \rightarrow \mathbb{R}$$

$K(\cdot, \cdot) \leftarrow \text{kernel}$

$$K(x, y) \in \mathbb{R}$$

$$t=x \Rightarrow K(x, \cdot) \in H \equiv K_x$$

parameterized
by x .

running
variable

$$K(\cdot, y) \in H \equiv K_y$$

running variable

→ The space of functions where dirac functional is continuous is known as RKHS.

RKHS : A subspace of L_2 where δ_t is continuous.

→ all functions need not be smooth.

eg:



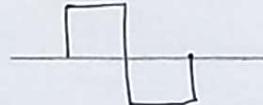
not smooth but \in RKHS

zero removed

→ functions with spikes are removed.

→ also few functions which are not continuous, are also part of space.

eg:



not continuous but \in RKHS.

i.e. enforcing atleast one type of continuity

(either right or left) are part of RKHS.

Proving uniqueness: By contradiction.

Let there are two Kernels K, K' .

If we show $K(\cdot, \cdot) = K'(\cdot, \cdot)$ for all points
then proof is done.

$$\|K(\cdot, \cdot) - K'(\cdot, \cdot)\|^2 = \langle K(\cdot, \cdot) - K'(\cdot, \cdot), K(\cdot, \cdot) - K'(\cdot, \cdot) \rangle$$

$\therefore [$ Kernel is unique to RKHS.]

$$= \langle K(\cdot, \cdot), K(\cdot, \cdot) \rangle + \langle K'(\cdot, \cdot), K'(\cdot, \cdot) \rangle - \langle K'(\cdot, \cdot), K(\cdot, \cdot) \rangle \\ - \langle K(\cdot, \cdot), K'(\cdot, \cdot) \rangle$$