

UNIT - 4. Information Theory:

Information :- The Information is a measure of uncertainty.

⇒ If the probability of occurrence is less than the information content (S) vice versa. i.e. the information and probability are inversely proportional to each other.

$$I(x_k) = \log \frac{1}{P(x_k)} \Rightarrow -\log P(x_k).$$

If base = 2, then it is called bits

base = 10, then it is called decimal

$$\therefore I(x_k) = \log_2 \frac{1}{P(x_k)}$$

Joint Probability of Information :-

Let us consider two events x_i & y_i having probabilities of occurrence $P(x_i)$ & $P(y_i)$ then $P(x_i, y_i)$ is called joint probability.

$$P(x_i, y_i) = P(x_i) \cdot P(y_i)$$

where $P(x_i)$, $P(y_i)$ are independent.

$$\text{So W.K.T } I(x_i, y_i) = \log_2 \frac{1}{P(x_i, y_i)}$$

$$\Rightarrow \log_2 \frac{1}{P(x_i) \cdot P(y_i)}$$

$$I(x_i, y_i) \Rightarrow \log_2 \frac{1}{P(x_i)} + \log_2 \frac{1}{P(y_i)}$$

$$\boxed{I(x_i, y_i) = I(x_i) + I(y_i)}$$

∴ Information content are also independent if probability of occurrence is independent.

Average information (g) Entropy (H) :-

⇒ Entropy is defined as the average information for individual message.

Let us assume m_1, m_2, \dots, m_M are different message having probability of occurrence P_1, P_2, \dots, P_m .

Let us assume that in a long interval of time 'L' message are generated where ($L \gg M$)

Let N_1 be the no of occurrence of m_1 , out of L message items

$$P_1 = \frac{N_1}{L} ; (N_1 = P_1 L)$$

$$P_2 = \frac{N_2}{L} ; (N_2 = P_2 L)$$

$$\vdots \quad \vdots$$

$$P_m = \frac{N_m}{L} ; (N_m = P_m L)$$

$$(N_1 + N_2 + \dots + N_m = L)$$

Let $I(t)$ is the total amount of information of (L) messages

$$\therefore I(t) = N_1 \log_2 \left(\frac{1}{P_1} \right) + N_2 \log_2 \left(\frac{1}{P_2} \right) + \dots + N_m \log_2 \left(\frac{1}{P_m} \right)$$

$$I(t) = P_1 L \log_2 \left(\frac{1}{P_1} \right) + P_2 L \log_2 \left(\frac{1}{P_2} \right) + \dots + P_m L \log_2 \left(\frac{1}{P_m} \right)$$

$$\frac{I(t)}{L} = \sum_{k=1}^m P_k \log_2 \left(\frac{1}{P_k} \right)$$

$$\therefore H = \frac{I(t)}{L} = \sum_{k=1}^m P_k \log_2 \left(\frac{1}{P_k} \right) \text{ & } - \sum_{k=1}^m P_k \log_2 (P_k)$$

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Case(1):- Show that if there is a single message the average information rate is equal to zero. i.e ($H=0$). if $P_k \leq 0.81$

$$H = \sum_{k=1}^m P_k \log_2 \left(\frac{1}{P_k} \right)$$

for single message $P=1$

$$\boxed{H = P \log_2 \left(\frac{1}{P} \right) \Rightarrow 0}$$

$$\therefore H = 1 \log_2 \left(\frac{1}{1} \right) = 0$$

$$\{ 0 \log_2 \frac{1}{0} = 0 \}$$

Case 2 :- Show that the maximum entropy occurs at $P = \frac{1}{2}$ whenever we are transmitting two msgs & draw P_H curve.

$$\text{Let } m_1 = P, m_2 = (1-P)$$

$$\therefore H = P_1 \log_2 \left(\frac{1}{P_1} \right) + P_2 \log_2 \left(\frac{1}{P_2} \right).$$

$$H \Rightarrow P \log_2 \left(\frac{1}{P} \right) + (1-P) \log_2 \left(\frac{1}{1-P} \right)$$

$$\Rightarrow - \left\{ P \log_2 P + (1-P) \log_2 (1-P) \right\}$$

$$\Rightarrow \frac{-1}{\ln 2} \left\{ P \ln P + (1-P) \ln (1-P) \right\}$$

$$\text{for max^n to be occur } \frac{dH}{dp} = 0$$

$$\frac{dH}{dp} \Rightarrow \frac{d}{dp} \left[\frac{-1}{\ln 2} \left\{ P \ln P + (1-P) \ln (1-P) \right\} \right]$$

$$\Rightarrow \frac{-1}{\ln 2} \left[P \cdot \frac{1}{P} + \ln P \cdot 1 + (1-P) \cdot \frac{1}{1-P} (-1) + \ln (1-P) (-1) \right]$$

$$\Rightarrow \frac{-1}{\ln 2} \left[1 + \ln P + (-1) + (-1) \ln (1-P) \right]$$

$$[\ln p - \ln(1-p)] = 0$$

$$\ln p = \ln(1-p)$$

$$p = 1-p$$

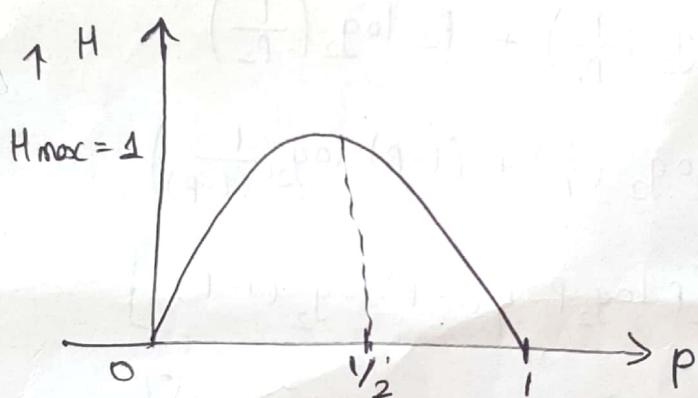
$$2p = 1$$

$$p = \frac{1}{2}$$

$$\therefore H = p \log \frac{1}{p} + (1-p) \log \left(\frac{1}{1-p} \right)$$

$$H_{\max} = \frac{1}{2} \log_2^2 + \frac{1}{2} \log_2^2 \Rightarrow 1$$

$H_{\max} = 1$ (hence max^m entropy occurs at $(P=\frac{1}{2})$)



Case(3) :- S.T If there are 'M' msg of equiprobable its entropy is given by $H = \log_2 M$.

Since we have 'M' msgs & all are equiprobable then

$$P_1 = P_2 = P_3 = \dots = P_M = \frac{1}{M}$$

$$H = \sum_{k=1}^M P_k \log_2 \frac{1}{P_k}$$

ab ab
ab ab

$$H = P_1 \log_2 \frac{1}{P_1} + P_2 \log_2 \frac{1}{P_2} + \dots + P_m \log_2 \frac{1}{P_m} \quad (3)$$

$$H = \frac{1}{M} \log_2 M + \frac{1}{M} \log_2 M + \dots + \frac{1}{M} \log_2 M$$

$$H = M \left(\frac{1}{M} \log_2 M \right)$$

$$\boxed{H = \log_2 M}$$

w.k.t $M = 2^N \rightarrow$ Sub in above

$$H = \log_2 2^N \Rightarrow N \log_2 2 \Rightarrow N \text{ bits/msg.}$$

$$\boxed{H = N \text{ bits/msg}}$$

Information Rate (R) :-

It is defined as the no. of bits transmitted per second if a source of messages transmitted at a rate of 'x' msg per second then the information rate is given by,

$$\boxed{R = x H \text{ bits/sec.}}$$

It is called max^m information rate.

The PGN & Observe

actual R $\leq 2 R_m$ (R_m is B)

because R_m

$$\therefore \boxed{R \leq 2 R_m}$$

⑧ An analog signal is band limited to 2kHz & sampled at Nyquist rate & quantized into 4 levels Q_1, Q_2, Q_3, Q_4 with probability $P_1 = \frac{1}{8}$, $P_2 = \frac{3}{8}$, $P_3 = \frac{3}{8}$, $P_4 = \frac{1}{8}$. Calculate information rate & entropy.

$$\text{N.K.T} \quad R = \gamma H.$$

$$B = 2\text{kHz}.$$

$$\gamma = 2B \Rightarrow 2(2\text{kHz}) \Rightarrow 4\text{kHz} = 4000 \text{ messages}.$$

$$\text{Calculate } H = \sum_{k=1}^M P_k \log_2 \left(\frac{1}{P_k}\right)$$

$$\Rightarrow \frac{1}{8} \log_2 8 + \frac{3}{8} \log_2 \frac{8}{3} + \frac{3}{8} \log_2 \frac{8}{3} + \frac{1}{8} \log_2 8$$

$$\Rightarrow \frac{1}{8} \log_2^3 + \frac{3}{8} \log_2^{\frac{8}{3}} + \frac{3}{8} \log_2^{\frac{8}{3}} + \frac{1}{8} \log_2^3$$

$$\Rightarrow \frac{3}{8} \log_2^2 + \frac{3}{8} \log_2^{\frac{8}{3}} + \frac{3}{8} \log_2^{\frac{8}{3}} + \frac{3}{8} \log_2^2$$

$$\Rightarrow 2 \text{ bits/msg}.$$

$$R = \gamma H$$

$$R = (4000) (\alpha)$$

$$R = 8000 \text{ bits/sec.}$$

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- ② A discrete system of source B emitting 8 msgs with probability $\frac{1}{2}, \frac{1}{8}, \frac{1}{8}, \frac{1}{16}, \frac{1}{16}, \frac{1}{16}, \frac{1}{32}, \frac{1}{32}$ calculate the avg information per msg, source & efficiency of the code.

$$\text{Ans} \quad H = \sum_{k=0}^7 p_k \log_2 \left(\frac{1}{p_k} \right)$$

$$\Rightarrow p_0 \log_2 \left(\frac{1}{p_0} \right) + p_1 \log_2 \left(\frac{1}{p_1} \right) + p_2 \log_2 \left(\frac{1}{p_2} \right) + p_3 \log_2 \left(\frac{1}{p_3} \right) + p_4 \log_2 \left(\frac{1}{p_4} \right) \\ + p_5 \log_2 \left(\frac{1}{p_5} \right) + p_6 \log_2 \left(\frac{1}{p_6} \right) + p_7 \log_2 \left(\frac{1}{p_7} \right)$$

$$\Rightarrow \frac{1}{2} \log_2 (2) + \frac{1}{8} \log_2 8 + \frac{1}{8} \log_2 8 + \frac{1}{16} \log_2 16 + \frac{1}{16} \log_2 16 + \frac{1}{16} \log_2 16 \\ + \frac{1}{32} \log_2 32 + \frac{1}{32} \log_2 32$$

$$\Rightarrow \frac{1}{2} + \frac{1}{8} \log_2^3 + \frac{1}{8} \log_2^3 + \frac{1}{16} \log_2^4 + \frac{1}{16} \log_2^4 + \frac{1}{16} \log_2^4 + \\ \frac{1}{32} \log_2^5 + \frac{1}{32} \log_2^5$$

$$\Rightarrow \frac{1}{2} + \frac{3}{8} + \frac{3}{8} + \frac{4}{16} + \frac{4}{16} + \frac{4}{16} + \frac{5}{32} + \frac{5}{32}$$

$$\Rightarrow \frac{37}{16} \text{ bits/msg}$$

$$\text{Avg. length of code } L = \frac{37}{16} \text{ bits}$$

$$\text{efficiency } \eta = \frac{H}{L} \Rightarrow \frac{37/16}{37/16} = 1 \text{ (81.100%)}.$$

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Joint Entropy :- The joint entropy represents entropy of joint occurrence of two or more events.

It is given as :-

$$H(x, y) = H(y, x) = \sum_{i=1}^m \sum_{j=1}^n p(x_i, y_j) \log_2 \frac{1}{p(x_i, y_j)}$$

here $H(y, x)$ represents entropy of joint occurrence of x & y .

Conditional Entropy :- The conditional entropy is also known as equivocation. The conditional entropy $H(y/x)$ represents uncertainty of x on average, when y is known.

likewise the conditional entropy $H(x/y)$ represents uncertainty of y on average when x is transmitted.

⇒ Thus conditional entropy indicates the information lost across the noisy channel. The mathematical equations are given by

$$H(x/y) = \sum_{i=1}^m \sum_{j=1}^n p(x_i, y_j) \log_2 \left(\frac{1}{p(x_i/y_j)} \right)$$

$$\& H(y/x) = \sum_{i=1}^m \sum_{j=1}^n p(x_i, y_j) \log_2 \left(\frac{1}{p(y_j/x_i)} \right)$$

③ Marginal Entropy :- when the entropy of individual event is calculated from joint probabilities of event is called marginal entropy.

→ In joint occurrence of x_i & y_j , from equation of probability of occurrence of $P(x_i)$ & $P(y_j)$. Then the entropies of x_i & y_j can be calculated i.e

$$H(x) = \sum_{i=1}^n P(x_i)$$

$$\text{Sub } P(x_i) = \sum_{j=1}^m P(x_i, y_j)$$

$$\text{Then } H(x) = \sum_{i=1}^n \sum_{j=1}^m P(x_i, y_j)$$

$$H(x) \Rightarrow \boxed{\sum_{i=1}^n \sum_{j=1}^m P(x_i, y_j)}$$

$$H(y) = \sum_{j=1}^m P(y_j)$$

putting $P(y_j)$ from above eq,

$$P(y_j) = \sum_{i=1}^n P(x_i, y_j)$$

$$H(y) = \boxed{\sum_{i=1}^n \sum_{j=1}^m P(x_i, y_j)}$$

Then entropies of $H(x)$ & $H(y)$ are called marginal entropies

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Relation between Joint Entropy & Conditional Entropy (8) Prove

that $H(x, y) = H(x/y) + H(y)$

Consider Joint entropy $H(x, y) = \sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \log_2 \left(\frac{1}{P(x_i, y_j)} \right)$

$$\Rightarrow -\sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \log_2 P(x_i, y_j) \quad \text{--- (1)}$$

from probability theory w.r.t

$$P(A/B) \cdot P(B) = P(AB)$$

$$P(x_i, y_j) = P(x_i/y_j) P(y_j) \quad \text{--- (2)}$$

Sub (2) in Eq (1)

$$H(x, y) = -\sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \log_2 P(x_i/y_j) P(y_j)$$

from logarithmic Properties

$$\log ab = \log a + \log b$$

$$H(x, y) = -\sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) [\log_2 P(x_i/y_j) + \log_2 P(y_j)] \quad \text{--- (3)}$$

Tence - from above Eq becomes

$$H(y, x) = -\sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \log_2 P(x_i/y_j)$$

$$\Rightarrow -\sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \log_2 P(y_j)$$

$$H(x, y) = \sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \left(\frac{1}{\log_2 P(x_i/y_j)} \right) + \sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \left(\frac{1}{\log_2 P(y_j)} \right)$$

$$\Rightarrow \sum_{i=1}^m \sum_{j=1}^m p(x_i, y_j) \log_2 \left(\frac{1}{p(x_i, y_j)} \right) + \sum_{j=1}^m \left\{ \sum_{i=1}^m p(x_i, y_j) \right\} \log_2 \frac{1}{p(y_j)}$$

The first term in above eqn is from standard probability theory.

$$H(X, Y) = H(X/Y) + \sum_{j=1}^m p(y_j) \log_2 \left(\frac{1}{p(y_j)} \right)$$

$$\left[\because \sum_{i=1}^m p(x_i, y_j) = p(y_j) \right]$$

$$H(X, Y) = H(X/Y) + H(Y)$$

Mutual Information :- The mutual information is defined as the

amount of information transferred when x_i is transmitted &

y_j is received. It is represented as

$$I(x_i, y_j) = \log \frac{p(x_i, y_j)}{p(x_i)}$$

here $I(x_i, y_j)$ = mutual if

$p(x_i, y_j)$ = conditional pr that x_i was transmitted & y_j is received

$p(x_i)$ = Probability of symbol x_i for transmission.

Average mutual Information :-

→ The average mutual information is represented by $I(X, Y)$. It is calculated in bit/Symbol. The avg mutual information is defined as amount of source information gained per received symbol. Here the avg mutual information is different from entropy.

It is given as

$$I(X:Y) = \sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) I(x_i, y_j)$$

Thus $I(x_i, y_j)$ is weighted by joint probabilities $P(x_i, y_j)$ over all possible joint events putting $I(x_i, y_j)$ in above eqn we get :-

$$I(X:Y) = \sum_{i=1}^m \sum_{j=1}^n P(x_i, y_j) \log_2 \frac{P(x_i, y_j)}{P(x_i)}$$

Properties of mutual information

- ① The mutual information channel is symmetric ie, $I(X:Y) = I(Y:X)$
- ② The mutual inf/l can be represented in terms of entropies of channel of channel i/p & o/p and condition entropies

$$I(X:Y) = H(X) - H(X|Y)$$

$$\Rightarrow H(Y) - H(Y|X)$$

here $H(X|Y)$ & $H(Y|X)$ are conditional entropies

- ③ The mutual information is always positive ie $I(X:Y) \geq 0$
- ④ The mutual information is related to the Joint entropy $H(X, Y)$ by following relation

$$I(X:Y) = H(X) + H(Y) - H(X, Y)$$

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Shannon's theorem channel Capacity :-

- ⇒ Given a source of "M" likely msgs with $M > 1$ which generating the information of a rate "R" & given channel capacity 'c'.
- Then if $R \leq c$ there existing a code technique, so that info of the source may be transmitted over a channel with P_e of error which is very small.
- A theorem which is complementary to Shannon's theorem and applied to a channel in which noise is gaussian in nature is known as Shannon's Noisy theorem.
- It states that channel capacity of a white band limited gaussian channel is given by

$$C = B \log_2 \left(1 + \frac{S}{N} \right) \text{ bits / sec.}$$

w.k.t $N = n \cdot B$

$$\therefore C = B \log_2 \left(1 + \frac{S}{nB} \right)$$

where B = Bandwidth

S = Signal power

N = Noise power Spectral density

C = channel capacity

Trade off b/w SNR & B.W:-

According to Shannon's capacity theorem $C = B \log_2 \left(1 + \frac{S}{N} \right)$

Case (i) :- if $N=0$ (ie. There is no noise in the channel).

$$\text{then } \frac{S}{N} = \infty$$

Such channel is called noiseless channel.

Capacity of such channel will be $C = B \log_2 (1 + \infty)$

$$C \Rightarrow \infty$$

Case (ii) :- when $B \cdot W \rightarrow \infty$, the channel capacity will be limited.

by B.W ↑, noise power also ↑'s & $(S/N) \downarrow$'s.

Hence B approaches ∞ & Capacity will be limited

$$C_{\infty} = \lim_{B \rightarrow \infty} C \Rightarrow 1.44 \frac{S}{N}$$

$$\textcircled{Q} \cdot \underline{S \cdot T} \text{ when } \underline{B \cdot W = \infty} \text{ then } c(\infty) = 1.44 \frac{S}{N}$$

$$\stackrel{?}{=} \text{W.K.T} \quad C = B \log_2 \left(1 + \frac{S}{N} \right)$$

$$\text{W.K.T} \quad N = nB$$

$$C = B \log_2 \left(1 + \frac{S}{nB} \right)$$

$$\text{mul \& dev by } \frac{S}{n}$$

$$C = \frac{S}{n} \cdot \frac{n}{S} B \log_2 \left(1 + \frac{S}{nB} \right)$$

$$C = \frac{S}{n} \cdot \frac{nB}{S} \log_2 \left(1 + \frac{S}{nB} \right)$$

$$C = \frac{S}{n} \log_2 \left(1 + \frac{S}{nB} \right)^{\frac{nB}{S}} \Rightarrow \frac{S}{n} \log_2 \left(1 + \frac{S}{nB} \right)^{\frac{1}{S/nB}}$$

$$\text{let } x = \frac{S}{nB}$$

then it is in form of $\log_2 (1+x)^{1/x}$

\therefore when $B \cdot W \rightarrow \infty$ then $x \rightarrow 0$

$$C_{\infty} = \frac{S}{n} \lim_{x \rightarrow 0} \log_2 \left(1 + \frac{S}{nB} \right)^{1/S/nB}$$

$$\Rightarrow \frac{S}{n} \lim_{x \rightarrow 0} \log_2 (1+x)^{1/x}$$

$$\Rightarrow \frac{S}{n} \log_2 e$$

$$\begin{array}{|c|} \hline \text{Let } (1+x)^{1/x} & \\ \hline \end{array}$$

$$\lim_{x \rightarrow 0} (1+x)^{1/x} = e$$

$$\boxed{C_{\infty} \Rightarrow 1.44 \frac{S}{n}}$$

\Rightarrow Shannon's limit.

Source Coding :-

- ⇒ Source coding techniques are used to remove the redundancy present in the data. It is achieved by assigning short code words for more probable msgs, whereas long code words for less probable msgs. Such coding is called Variable length coding.
- ⇒ In binary encoding all msgs are assigned same length code irrespective of their probability of occurrence.
- ⇒ Codeword assigned to each msg is unique & the system that generate source coding is called source encoder.
- ⇒ Coding efficiency :- $(\eta) = \frac{H}{\bar{n}}$

where H is entropy.

\bar{n} is avg no of bits in cod word of msg

$$\bar{n} = \sum_{k=0}^{L-1} p_k n_k$$

Code redundancy :- It is the measure of bits in the encoded msg sequence. It is given as

$$\begin{aligned} \text{Redundancy } (V) &= 1 - \text{code efficiency} \\ &\Rightarrow 1 - \eta \end{aligned}$$

There are two types of coding :-

- ① Shannon - Fanor coding
- ② Huffman coding.

Shannon Fano Coding:- It is an algorithm which is a Variable length coding in which each msg is represented in terms of Variable bits per msg.

→ It reduce the redundancy and we can represent the msg with few no of bits so that B.W can be reduced & noise immunity can be improved.

Steps:-

- ① The messages are arranged according to their decreasing probabilities
- ② partition the msg such that probabilities ~~sum~~ in both the partitions are almost equal.
- ③ The msgs in upper partition are assigned to bit "0" & lower partition assigned to bit "1".
- ④ Each partition is further subdivided into new partitions following same rule.
- ⑤ Partitioned is stopped when there is only one msg in partition.
- ⑥ Find out the code efficiency by Shannon fano code for given ~~prob~~ messages.

message	m_1	m_2	m_3	m_4	m_5	m_6	m_7	m_8
$P(m_i)$	$\frac{1}{2}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{1}{16}$	$\frac{1}{16}$	$\frac{1}{32}$	$\frac{1}{32}$

message	Probability	1	2	3	4	5	no 8 bits / msg
m_1	$\frac{1}{2}$	0					1
m_2	$\frac{1}{8}$	1	0	0			3
m_3	$\frac{1}{8}$	1	0	1			3
m_4	$\frac{1}{16}$	1	1	0	0		4
m_5	$\frac{1}{16}$	1	1	0	1		4
m_6	$\frac{1}{16}$	1	1	1	0		4
m_7	$\frac{1}{32}$	1	1	1	1	0	5
m_8	$\frac{1}{32}$	1	1	1	1	1	5

$$\eta = \frac{H}{L} \times 100\%$$

$$H = \sum_{k=1}^8 P_k \log_2 \frac{1}{P_k} = P_1 \log_2 \frac{1}{P_1} + P_2 \log_2 \frac{1}{P_2} + \dots + P_8 \log_2 \frac{1}{P_8}$$

$$\Rightarrow \frac{1}{2} \log_2 \frac{1}{\frac{1}{2}} + \frac{1}{8} \log_2 \frac{1}{\frac{1}{8}} + \frac{1}{8} \log_2 \frac{1}{\frac{1}{8}} + \frac{1}{16} \log_2 \frac{1}{\frac{1}{16}} + \frac{1}{16} \log_2 \frac{1}{\frac{1}{16}} + \frac{1}{16} \log_2 \frac{1}{\frac{1}{16}}$$

$$+ \frac{1}{32} \log_2 \frac{1}{\frac{1}{32}} + \frac{1}{32} \log_2 \frac{1}{\frac{1}{32}}$$

$$\Rightarrow \frac{37}{16}$$

$$L \Rightarrow \sum_{i=1}^8 P_k x_i(L) \Rightarrow P_1 x_1 + P_2 x_2 + P_3 x_3 + \dots + P_8 x_8$$

$$\Rightarrow \frac{1}{2} \times 1 + \frac{1}{8} \times 3 + \frac{1}{8} \times 3 + \frac{1}{16} \times 4 + \frac{1}{16} \times 4 + \frac{1}{32} \times 5 + \frac{1}{32} \times 5$$

$$\Rightarrow \frac{37}{16}$$

$$\eta = \frac{37/16}{37/16} \Rightarrow 100\%$$

② Find the code efficiency by shannon fanon coding for 4 quantization levels.
 $\theta_1, \theta_2, \theta_3, \theta_4$ having probability $\frac{1}{8}, \frac{3}{8}, \frac{3}{8}, \frac{1}{8}$.

S

msg	Probability	1	2	3	bits/msg
m_1	$\frac{3}{8}$	0			1
m_2	$\frac{3}{8}$	1	0		2
m_3	$\frac{1}{8}$	1	1	0	3
m_4	$\frac{1}{8}$	1	1	1	3

$$\eta \% = \frac{H}{L} \%$$

$$H = \sum_{k=1}^4 p_k \log_2 \left(\frac{1}{p_k} \right)$$

$$\Rightarrow \frac{3}{8} \log_2 \frac{1}{\frac{3}{8}} + \frac{3}{8} \log_2 \frac{1}{\frac{3}{8}} + \frac{1}{8} \log_2 \frac{1}{\frac{1}{8}} + \frac{1}{8} \log_2 \frac{1}{\frac{1}{8}}$$

$$\Rightarrow 1.8112$$

$$L = \sum_{i=1}^4 p_k L(x) \Rightarrow p_1 b_1 + p_2 L_2 + p_3 L_3 + p_4 L_4$$

$$\Rightarrow \frac{3}{8} \times 1 + \frac{3}{8} \times 2 + \frac{1}{8} \times 3 + \frac{1}{8} \times 3 = \frac{15}{8}$$

$$\% \eta = \frac{H}{L} \times 100 \Rightarrow \frac{1.8112}{15/8} \times 100$$

$$\Rightarrow 96.6 \%$$

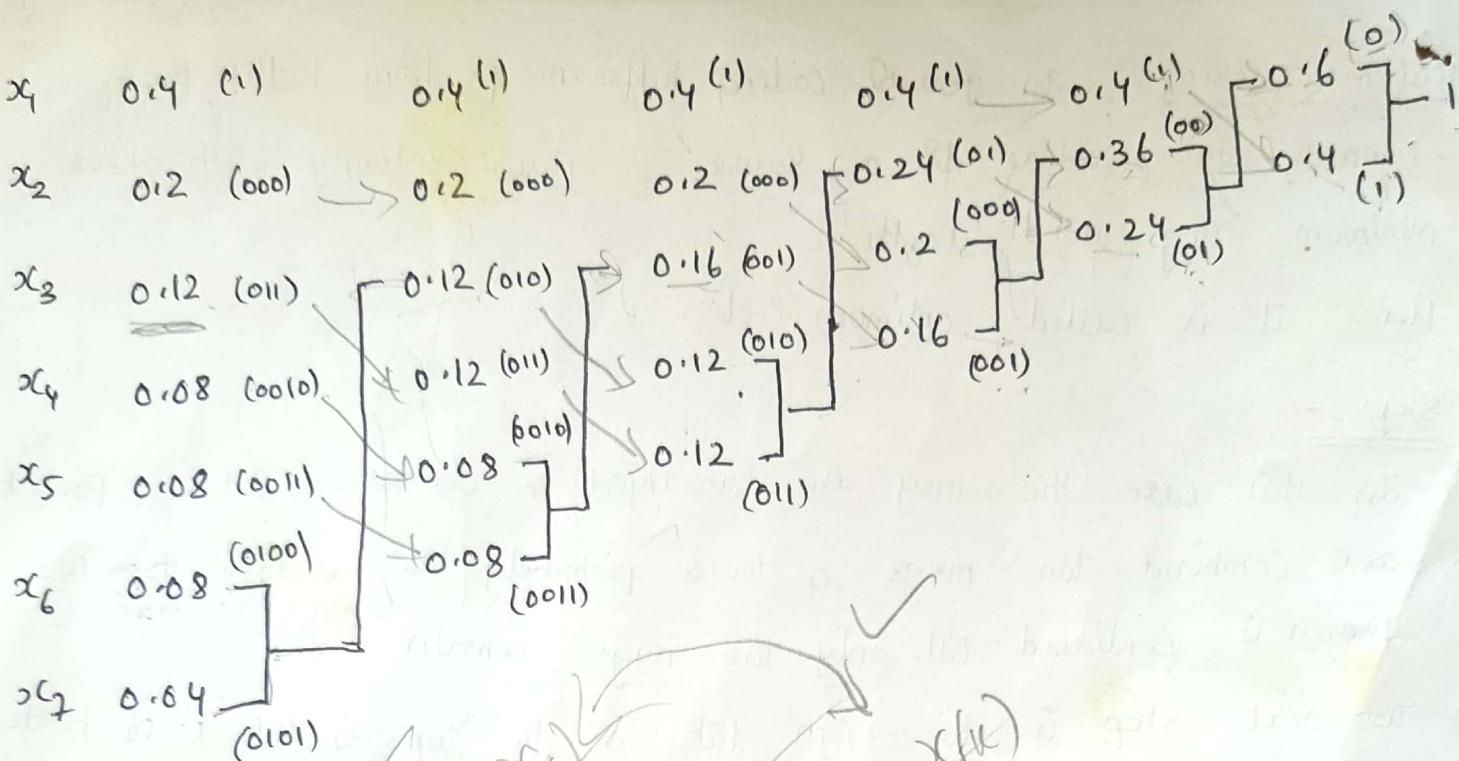
Huffman coding :- In general coding huffman perform better than shannon fano coding for a source of given entropy which gives minimum avg word length.

Hence it is called optimum code.

Steps:-

- ① In this case the msg are arranged in decreasing order & probability are combined two msgs & lower probability is assigned then this process is continued till only two msgs remain
- ② The next step is to assign bit '0' to top & bit '1' to bottom probabilities
- ③ The encoded msg is represented by the bits goes to starting from the final reduction.
- ④ Construct a huffman code & calculate the efficiency for given msg :-

msg	x_1	x_2	x_3	x_4	x_5	x_6	x_7
$p(x)$	0.4	0.12	0.12	0.08	0.08	0.08	0.04



Symbol	Probability	Code word	Length
x_1	0.4	1	1
x_2	0.2	000	3
x_3	0.12	011	3
x_4	0.08	0010	4
x_5	0.08	0011	4
x_6	0.08	0100	4
x_7	0.04	0101	4

$$\% \eta = \frac{H}{E} \times 100\%$$

$$L = \sum_{k=1}^7 p_k L(x) .$$

$$\Rightarrow 0.4 \times 1 + 0.2 \times 3 + 0.12 \times 3 + 0.08 \times 4 + 0.08 \times 4 + 0.08 \times 4 + 0.04 \times$$

$$L \Rightarrow Q \cdot 4g$$

$$H = \sum_{k=1}^7 P_k \log_2 \frac{1}{P_k}$$

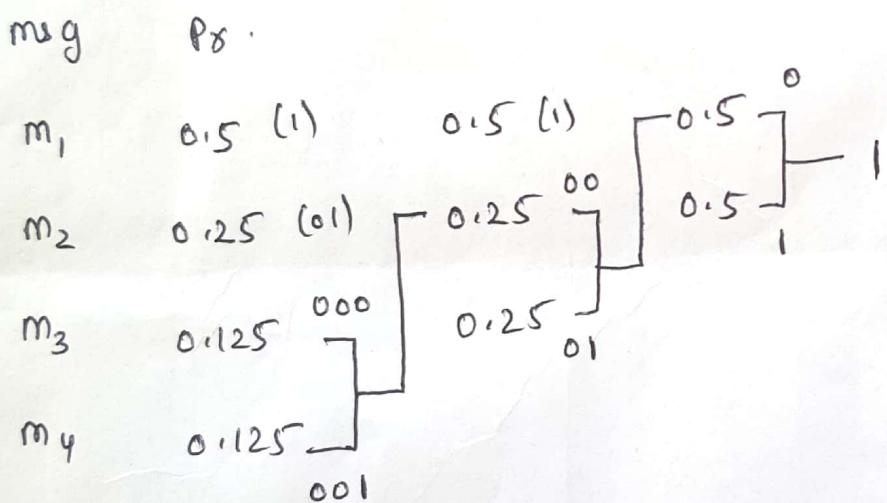
$$\Rightarrow 0.4 \log_2 \frac{1}{0.4} + 0.2 \log_2 \frac{1}{0.2} + 0.12 \log_2 \frac{1}{0.12} + 0.08 \log_2 \frac{1}{0.08} \\ + 0.08 \log_2 \frac{1}{0.08} + 0.08 \log_2 \frac{1}{0.08} + 0.04 \log_2 \frac{1}{0.04}$$

$$\Rightarrow 2.48$$

$$\eta = \frac{H}{L} \times 100 \Rightarrow \frac{2.42}{2.48} \times 100\% \Rightarrow 97.6\%$$

② There are 4 msgs & its probability are

msg	M ₁	M ₂	M ₃	M ₄
P _x (msg)	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{8}$



msg	Prob codes	code word	length
M ₁	0.5	1	1
M ₂	0.25	01	2
M ₃	0.125	000	3
M ₄	0.125	001	3

$$\% \text{Eff} = \frac{H}{L}$$

$$H = P_1 \log_2 \frac{1}{P_1} + P_2 \log_2 \frac{1}{P_2} + P_3 \log_2 \frac{1}{P_3} + P_4 \log_2 \frac{1}{P_4}$$

$$\Rightarrow 0.5 \log_2 \frac{1}{0.5} + 0.25 \log_2 \frac{1}{0.25} + 0.125 \log_2 \frac{1}{0.125} + 0.125 \log_2 \frac{1}{0.125}$$

\Rightarrow

$$L = \sum_{i=1}^4 P_k L(x_k)$$

$$\Rightarrow 0.5(1) + 0.25(2) + 0.125(3) + 0.125(3)$$

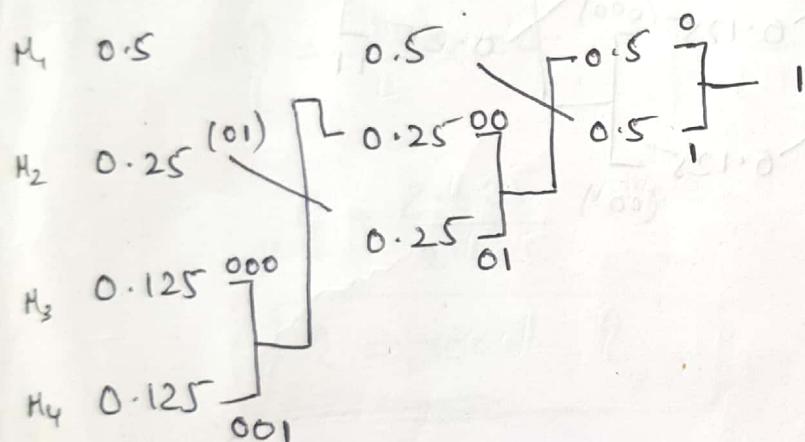
\Rightarrow

Huffman Coding :- It is generally used for data compression.

① There are 4 messages & their probabilities are

Msg(M)	M ₁	M ₂	M ₃	M ₄
Probability P(m)	$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{8}$
	(0.5)	(0.25)	(0.125)	(0.125)

≡ Step ① Arrange in descending order.



Codes

$$M_1 \rightarrow (1)$$

$$M_2 \rightarrow (01)$$

$$M_3 \rightarrow (000)$$

$$M_4 \rightarrow (001)$$

Q) Consider a source with 7 msgs having probabilities 0.25, 0.25, 0.125, 0.125, 0.125, 0.0625, 0.0625. Find avg code length, Entropy & efficiency?

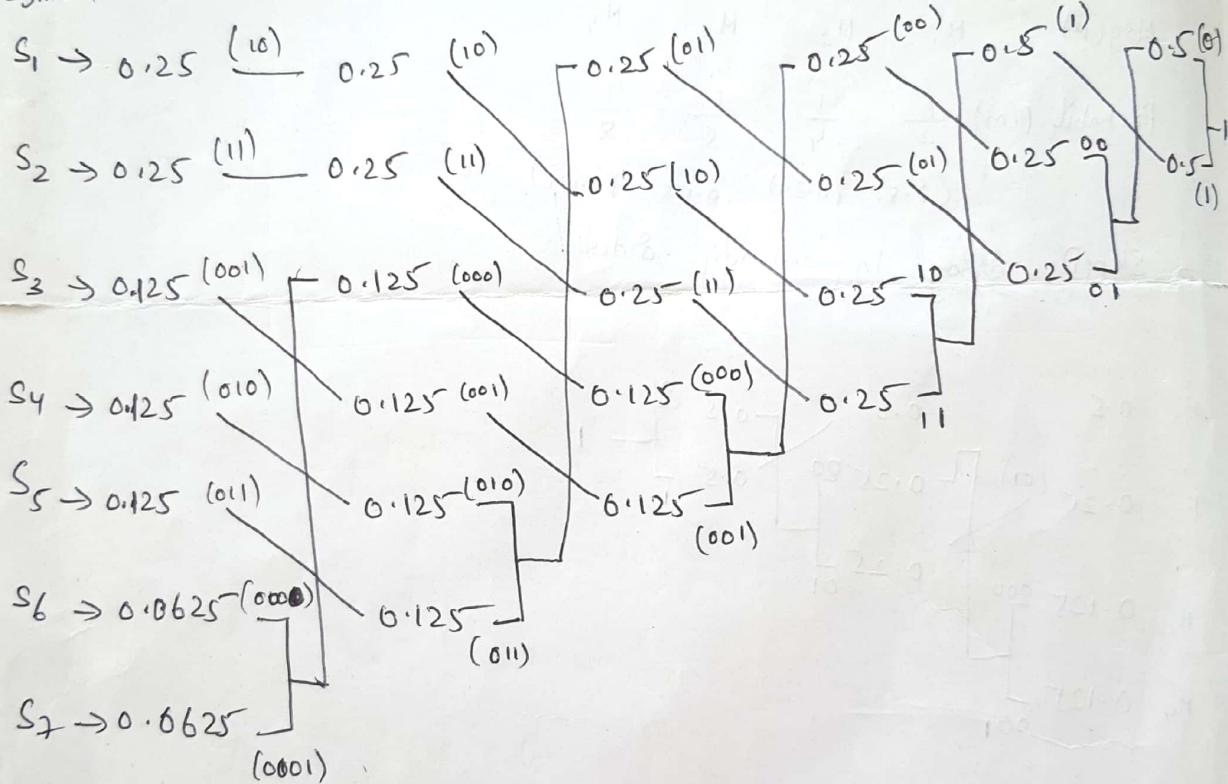
Sol). ① Write probabilities in descending order.

$$② \text{ Avg code length } L = \sum_{k=1}^K P_k L_k$$

$$③ \text{ Entropy } H(x) = \sum_{k=1}^K P_k \log_2 P_k$$

$$④ \text{ Efficiency } \eta = \frac{H}{L} \times 100$$

Symbol



Symbol	Prob	Codeword	Length
S_1	0.25	10	2
S_2	0.25	11	2
S_3	0.125	(001)	3
S_4	0.125	(010)	3
S_5	0.125	(011)	3
S_6	0.0625	(0000)	4
S_7	0.0625	(0001)	4

$$L = \sum_{k=1}^n P_k L_k$$

$$L = P_1 L_1 + P_2 L_2 + P_3 L_3 + P_4 L_4 + P_5 L_5 + P_6 L_6 + P_7 L_7$$

$$\Rightarrow 0.25 \times 2 + 0.25 \times 2 + 0.125 \times 3 + 0.125 \times 3 + 0.125 \times 3 + \\ 0.0625 \times 4 + 0.0625 \times 4$$

$\Rightarrow 2.625$ bits / symbol

$$H(X) = - \left(\sum_{k=1}^n P_k \log_2 P_k \right)$$

$$\Rightarrow P_1 \log_2 P_1 + P_2 \log_2 P_2 + P_3 \log_2 P_3 + P_4 \log_2 P_4 + P_5 \log_2 P_5 \\ + P_6 \log_2 P_6 + P_7 \log_2 P_7$$

$$\Rightarrow (0.25 \log 0.25)_2 + (0.125 \log 0.125)_3 + (0.0625 \log 0.0625)_4$$

$\Rightarrow 2.625$ bits / symbol

$$\eta = \frac{H}{L} \times 100$$

$$\eta = \frac{2.625}{2.625} \times 100$$

$$\boxed{\eta \Rightarrow 100 \text{ or } 2}$$

① Apply shannon fano code for following msg:-

$$(x) = x_1 \ x_2 \ x_3 \ x_4 \ x_5 \ x_6$$

$$(P_x) = 0.30 \ 0.25 \ 0.15 \ 0.12 \ 0.08 \ 0.10$$

S ① Arrange all P_x in descending order.

[x]	[P_x]	s_1	Reduced s_2	Structure s_3	Code word	bth
x_1	0.30	0	0		00	2
x_2	0.25	0	1		01	2
x_3	0.15	1	0	0	100	3
x_4	0.12	1	0	1	101	3
x_6	0.10	1	1	0	110	3
x_5	0.08	1	1	1	111	3

$$\text{entropy } H(x) = \sum_{i=1}^6 P_i \cdot I_i = \sum_{i=1}^6 P_i \log \frac{1}{P_i}$$

$$\Rightarrow 0.30 \log \frac{1}{0.30} + 0.25 \log \frac{1}{0.25} + 0.15 \log \frac{1}{0.15} + \dots - 0.08 \log \frac{1}{0.08}$$

$$\Rightarrow 2.418 \text{ bds}$$

$$\text{Avg} = \sum_{i=1}^6 P_i \times L_i = 0.30(2) + 0.25(2) + 0.15 \times 3 + 0.12 \times 3 + 0.10 \times 3 + 0.08 \times 3$$

$$\Rightarrow 2.45 \text{ bds}$$

$$\eta = \frac{H(x)}{\text{Avg}} = \frac{2.418}{2.45} = 0.9869$$

$$\eta = 98.69\%$$

$$(\text{Ordinary}) \quad \eta = 1 - \eta = 1 - 0.9869 = 0.0131 \text{ or } 1.31\%$$

Roll No.

msg	P_x	s_1	s_2	s_3	s_4	s_5 no of bits	Code
m_1	$\frac{1}{2}$	0				1	0
m_2	$\frac{1}{8}$	1	0	0		3	100
m_3	$\frac{1}{8}$	1	0	1		3	101
m_4	$\frac{1}{16}$	1	1	0	0	4	1100
m_5	$\frac{1}{16}$	1	1	0	1	4	1101
m_6	$\frac{1}{32}$	1	1	1	0 0	5	11100

7.15

④ A Discrete Memoryless Source has five symbols x_1, x_2, x_3, x_4, x_5 with $P(x_1) = 0.4, P(x_2) = 0.19, P(x_3) = 0.16, P(x_4) = 0.15$ & $P(x_5) = 0.1$?

- ✓ ① Construct a Shannon Fano code for x , calculate efficiency.
- ② " Huffman Code ?

⑤ Determine Huffman code for following msg with their Probabilities

x_1	x_2	x_3	x_4	x_5	x_6	x_7
0.05	0.15	0.2	0.05	0.15	0.3	0.1

PULSE MODULATION

Introduction:

Pulse Modulation

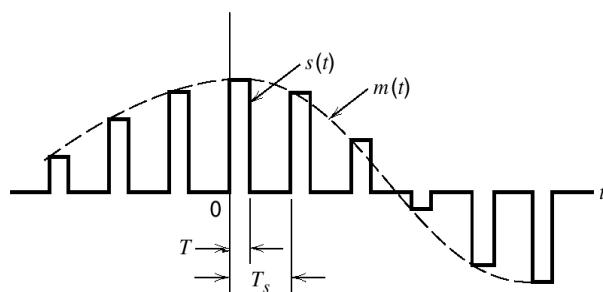
- Carrier is a train of pulses
- Example: Pulse Amplitude Modulation (PAM), Pulse width modulation (PWM) , Pulse Position Modulation (PPM)

Types of Pulse Modulation:

- The immediate result of sampling is a pulse-amplitude modulation (PAM) signal
- PAM is an analog scheme in which the amplitude of the pulse is proportional to the amplitude of the signal at the instant of sampling
- Another analog pulse-forming technique is known as **pulse-duration modulation (PDM)**. This is also known as **pulse-width modulation (PWM)**
- **Pulse-position modulation** is closely related to PDM

Pulse Amplitude Modulation:

In PAM, amplitude of pulses is varied in accordance with instantaneous value of modulating signal.



PAM Generation:

The carrier is in the form of narrow pulses having frequency f_c . The uniform sampling takes place in multiplier to generate PAM signal. Samples are placed T_s sec away from each other.

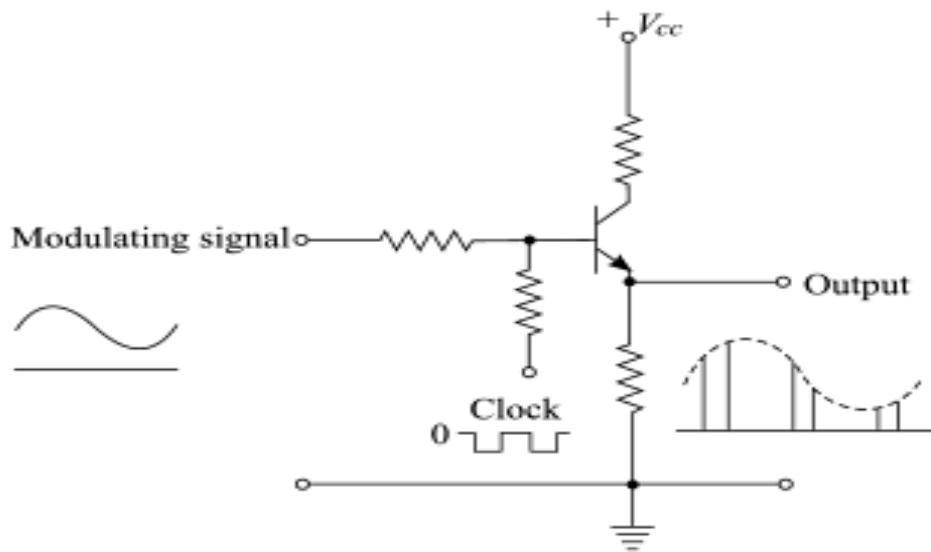


Fig.12. PAM Modulator

- The circuit is simple emitter follower.
- In the absence of the clock signal, the output follows input.
- The modulating signal is applied as the input signal.
- Another input to the base of the transistor is the clock signal.
- The frequency of the clock signal is made equal to the desired carrier pulse train frequency.
- The amplitude of the clock signal is chosen such that the high level is at ground level(0v) and low level at some negative voltage sufficient to bring the transistor in cutoff region.
- When clock is high, circuit operates as emitter follower and the output follows the input modulating signal.
- When clock signal is low, transistor is cutoff and output is zero.
- Thus the output is the desired PAM signal.

PAM Demodulator:

- The PAM demodulator circuit which is just an envelope detector followed by a second order op-amp low pass filter (to have good filtering characteristics) is as shown below

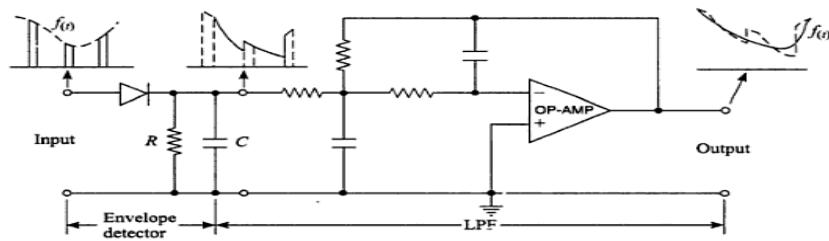
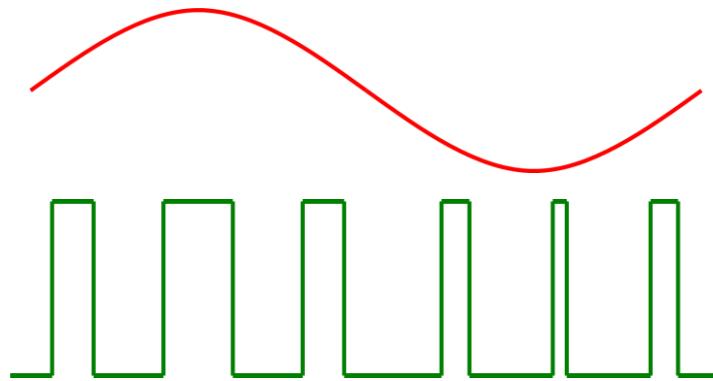


Fig.13. PAM Demodulator

Pulse Width Modulation:

- In this type, the amplitude is maintained constant but the width of each pulse is varied in accordance with instantaneous value of the analog signal.



- In PWM information is contained in width variation. This is similar to FM.
- In pulse width modulation (PWM), the width of each pulse is made directly proportional to the amplitude of the information signal.

Pulse Position Modulation:

- In this type, the sampled waveform has fixed amplitude and width whereas the position of each pulse is varied as per instantaneous value of the analog signal.
- PPM signal is further modification of a PWM signal.

PPM & PWM Modulator:

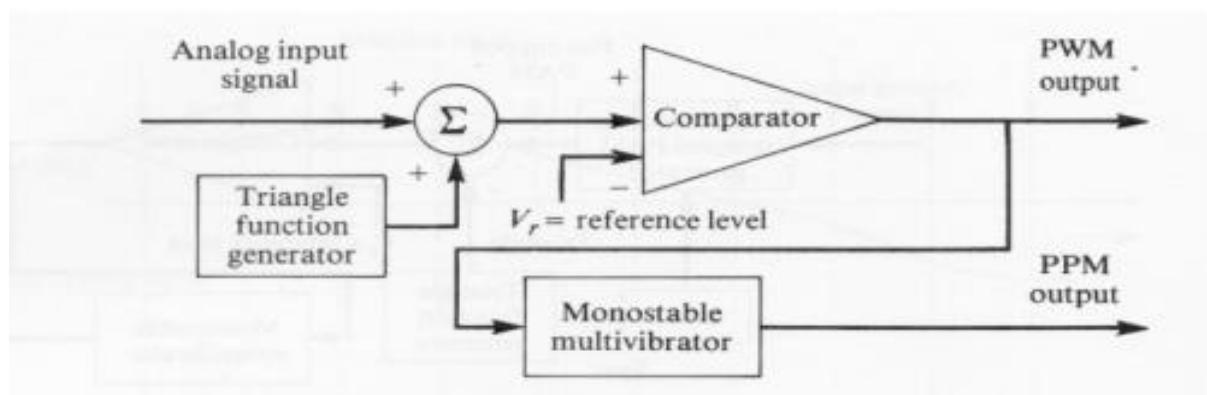


Fig.14. PWM & PPM Modulator

- The PPM signal can be generated from PWM signal.
- The PWM pulses obtained at the comparator output are applied to a mono stable multi vibrator which is negative edge triggered.

- Hence for each trailing edge of PWM signal, the monostable output goes high. It remains high for a fixed time decided by its RC components.
- Thus as the trailing edges of the PWM signal keeps shifting in proportion with the modulating signal, the PPM pulses also keep shifting.
- Therefore all the PPM pulses have the same amplitude and width. The information is conveyed via changing position of pulses.

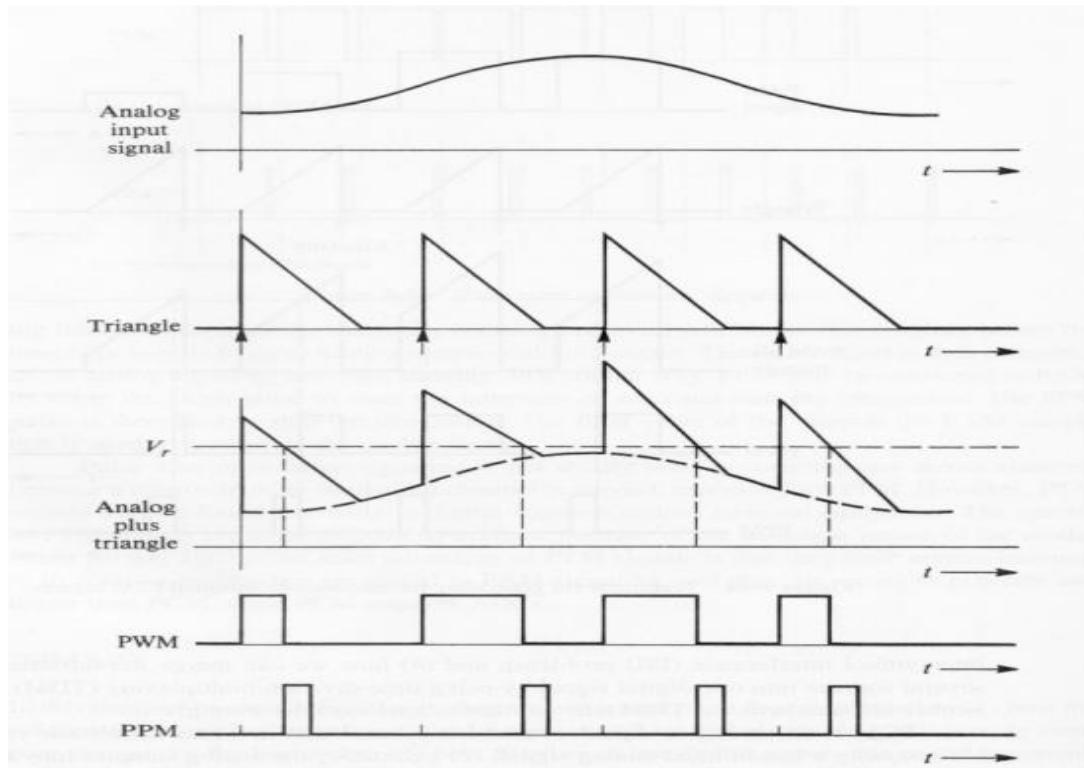


Fig.15. PWM & PPM Modulation waveforms

PWM Demodulator:

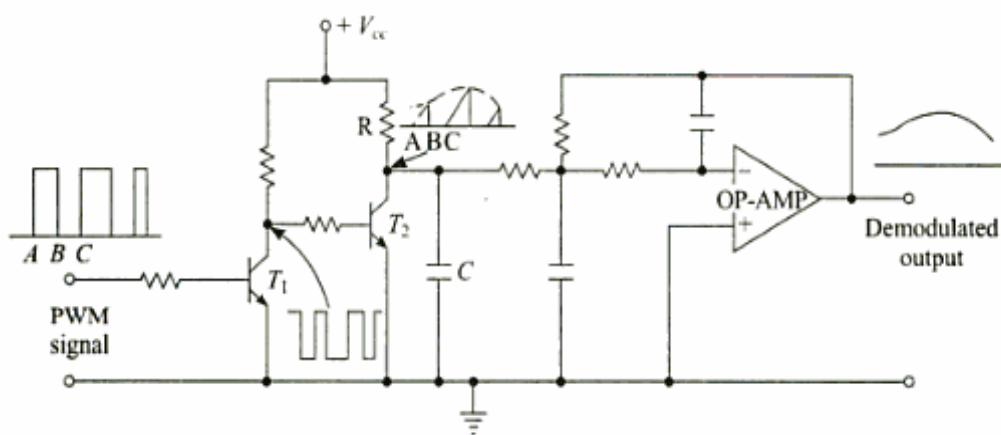


Fig.16. PWM Demodulator

- Transistor T1 works as an inverter.
- During time interval A-B when the PWM signal is high the input to transistor T2 is low.
- Therefore, during this time interval T2 is cut-off and capacitor C is charged through an R-C combination.
- During time interval B-C when PWM signal is low, the input to transistor T2 is high, and it gets saturated.
- The capacitor C discharges rapidly through T2. The collector voltage of T2 during B-C is low.
- Thus, the waveform at the collector of T2 is similar to saw-tooth waveform whose envelope is the modulating signal.
- Passing it through 2nd order op-amp Low Pass Filter, gives demodulated signal.

PPM Demodulator:

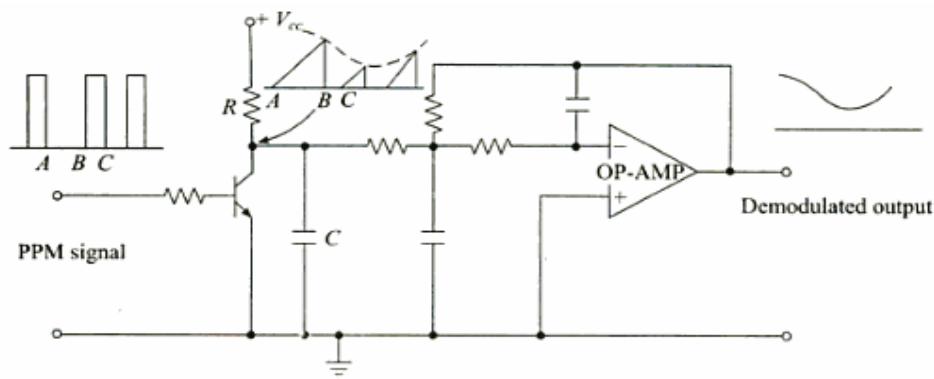
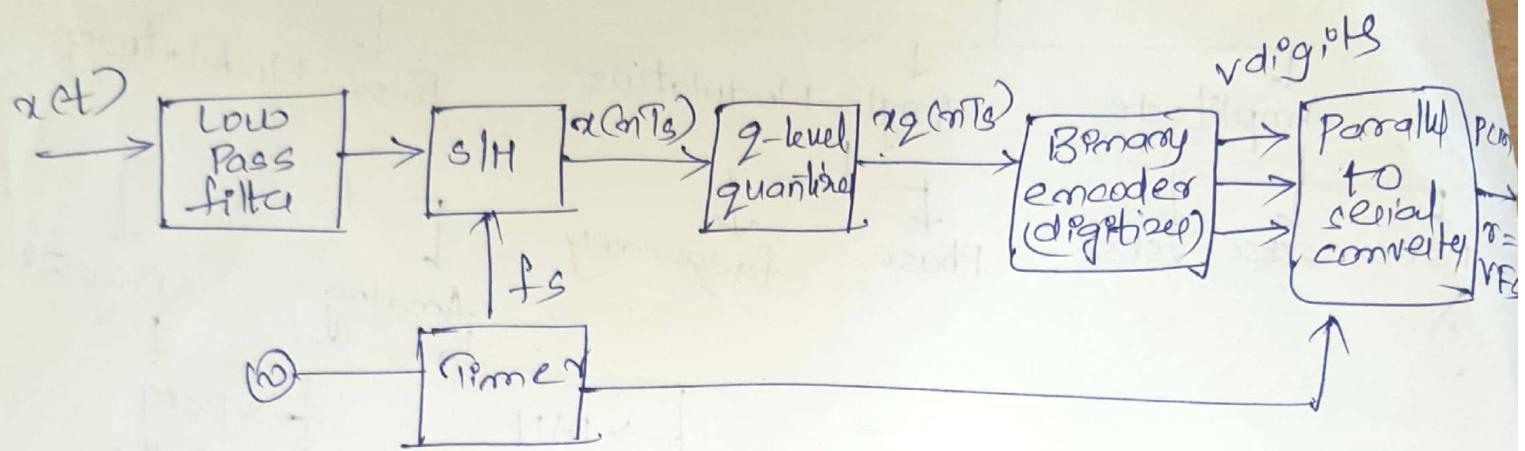


Fig.17. PPM Demodulator

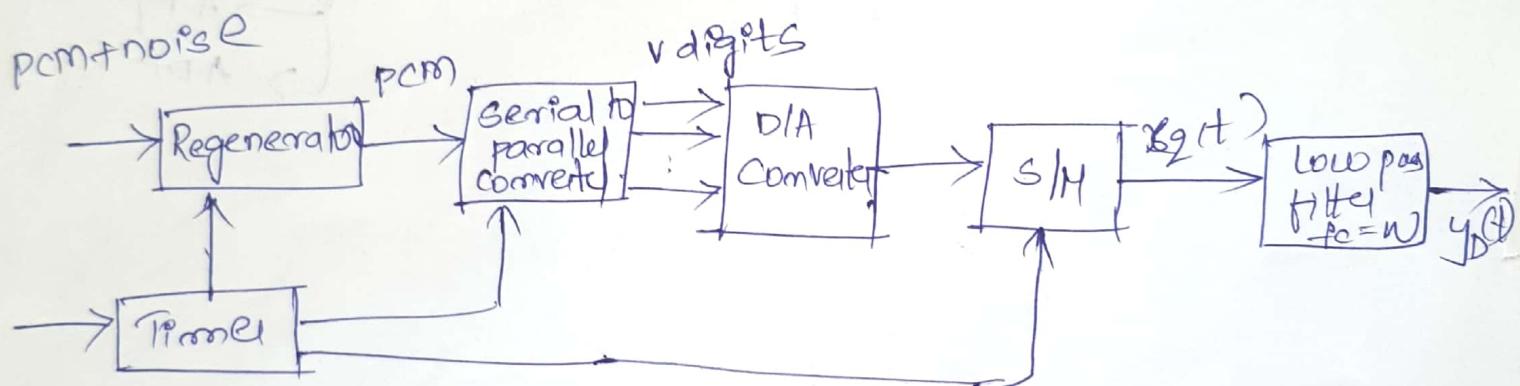
- The gaps between the pulses of a PPM signal contain the information regarding the modulating signal.
- During gap A-B between the pulses the transistor is cut-off and the capacitor C gets charged through R-C combination.
- During the pulse duration B-C the capacitor discharges through transistor and the collector voltage becomes low.
- Thus, waveform across collector is saw-tooth waveform whose envelope is the modulating signal.
- Passing it through 2nd order op-amp Low Pass Filter, gives demodulated signal.

PCM (Pulse Code Modulation)

PCM Generator



PCM Receiver



PCM

Transmitter Explanation

Low Pass filter: This low pass filter blocks all the frequency components above 'W' Hz. Thus $x(t)$ is bandlimited to 'W' Hz.

Sample and Hold circuit: The sampled and hold circuit then samples this signal at the rate of f_s . Sampling frequency f_s is selected sufficiently above Nyquist rate, i.e., $f_s \geq 2W$ and output of sample and hold is called $x(nT_s)$. $x(nT_s)$ is discrete in time and continuous in Amplitude.

2-level quantizer: A 2 level quantiser compares the input $x(nT_s)$ with its fixed digital levels. It assigns any one of the digital level to $x(nT_s)$ which results in minimum distortion or error. This error is called quantisation error. This output of quantizer is a digital level called $q_2(nT_s)$.

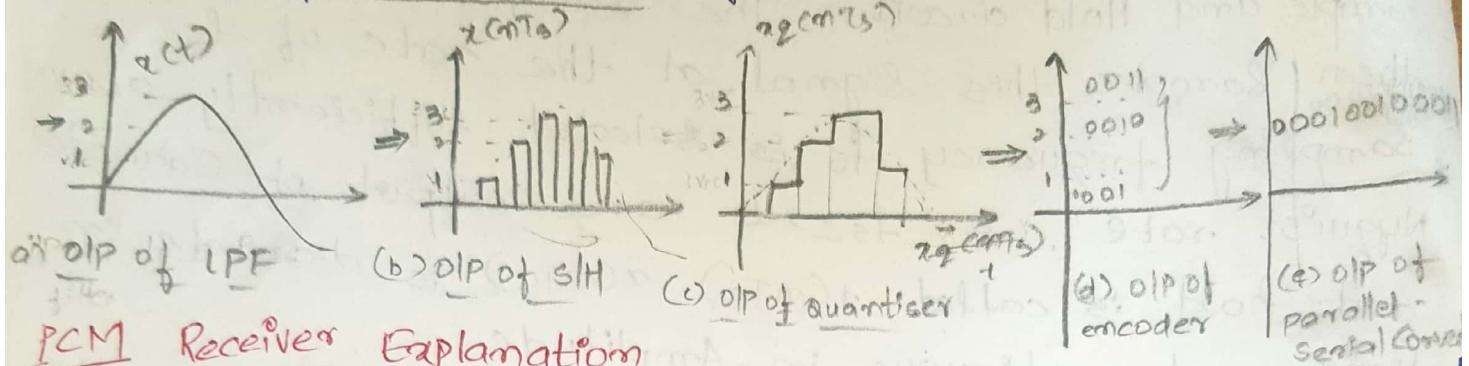
Binary Encoder: the quantised signal level $q_2(nT_s)$ is given to binary encoder. This encoder converts parallel signal to 'v' digits binary word. Thus $q_2(nT_s)$ is converted to 'v' binary bits. The encoder is called digitizer.

Parallel to Serial Converter:

It is not possible to transmit each bit of the binary word separately on transmission line. Therefore 'v' binary bits are converted to serial bit stream to generate single baseband signal. In a parallel to serial converter, normally a shift register do this job. Thus the output of Pcm generator is thus a single baseband of binary bits.

→ Thus, an oscillator/timer generates the clock pulses for sample and hold and parallel to serial converter.

→ Hence, sample and hold, quantizer and encoder combinedly form Analog to digital converter.



PCM Receiver Explanation

Regenerator:- It is impossible to reconstruct exact signal at ADC because of permanent quantisation introduced during quantisation at the transmitted. So, noise can be reduced by taking the step size, and this can be done with the help of regenerative circuit.

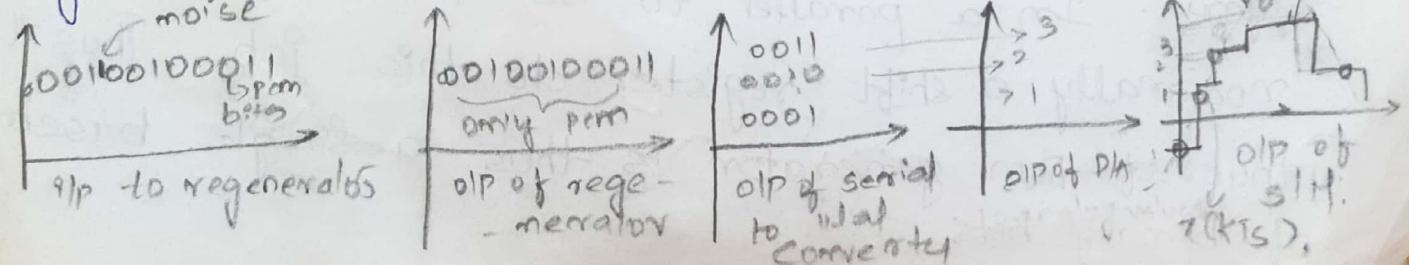
Serial - Parallel Converter:- The Serially transmitted bits are converted to parallel to reconstruct the samples.

Decoder:- Decoder converts serial binary digit to

D/A converter:- It converts the parallelly transmitted binary bits to the analog values

SH → Sample and hold will produce the sampled output and also produces the samples quantised output. Thus SH performs both sampling and quantisation o/p in the Receiver.

Low Pass filter:- Low Pass filter will reconstruct the original bandwidth w signal from SH O/P.



QUANTISATION:-

The input Sample value is quantized to nearest digital level. The difference between two quantisation levels is called step size(δ).

Quantisation:-

↓
Uniform linear
step size is constant
over the amplitude
range

↓
Midread Quantiser Midiser Quantiser Biased.

↓ Non Quantisation Non linear
step size is varying.

a) Midread Quantiser:-

The transfer characteristic of the midread quantizer as shown in figure.

$\delta \rightarrow$ step size.

For, $-8\delta/2 \leq x(\text{mTS}) \leq 8\delta/2$,

$$xg(\text{mTS}) = 0$$

$$8\delta/2 \leq x(\text{mTS}) \leq 38\delta/2, xg(\text{mTS}) = 8$$

It is called midread

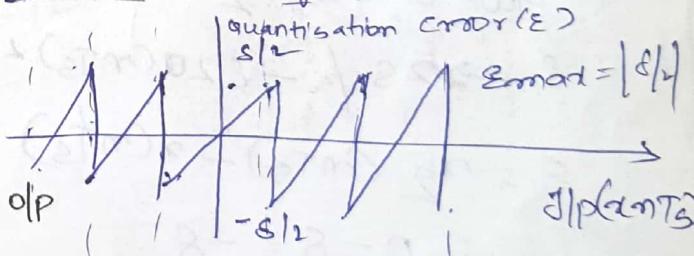
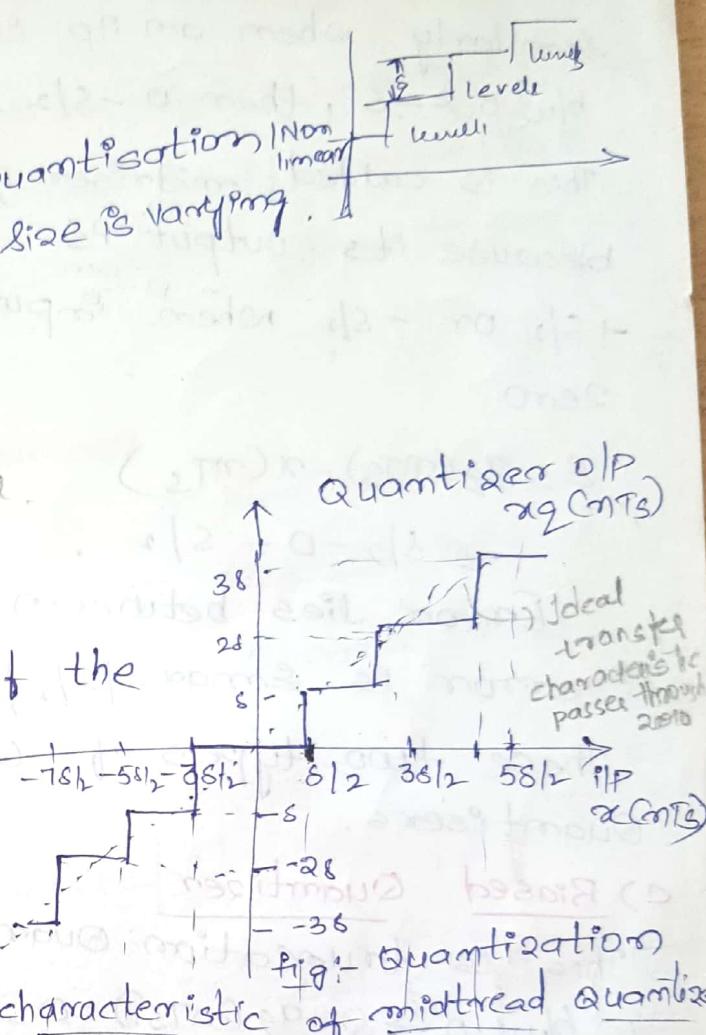
Quantizer because Quantizer o/p

is zero when $x(\text{mTS}) = 0$.

$$\epsilon = xg(\text{mTS}) - x(\text{mTS})$$

So, error lies between $-8\delta/2 \leq \epsilon \leq 8\delta/2$. So, the

max. quantisation error will be given as $E_{\text{max}} = \left| \frac{8}{2} \right|$



(b) Midriser Quantiser

When the input is between 0 & δ , the dlp is $\pm \frac{\delta}{2}$,

similarly when am dlp is

btw $0 \& -\delta$, then $0 - \frac{\delta}{2}$.

This is called midriser quantiser

because its output is either

$+\frac{\delta}{2}$ or $-\frac{\delta}{2}$ when input is

zero

$$\epsilon = q_2(m_{TS}) - x(m_{TS})$$

$$= \frac{\delta}{2} - 0 = \frac{\delta}{2}.$$

Error lies between $0 - \frac{\delta}{2} + \frac{\delta}{2}$, the max. quantisation error is $\epsilon_{max} = \frac{\delta}{2}$.

These two types of Quantisers are rounding Quantisers.

c) Biased Quantizer

This is truncation Quantiser

btw $0 \rightarrow \delta$, $q_2(m_{TS}) = 0$,

$\delta \rightarrow 2\delta$ / -2δ , $q_2(m_{TS}) = 0$, $q_2(m_{TS}) = -\delta$

$$\epsilon = q_2(m_{TS}) - x(m_{TS})$$

$$= 0 - \delta = -\delta.$$

Hence, max Quantisation

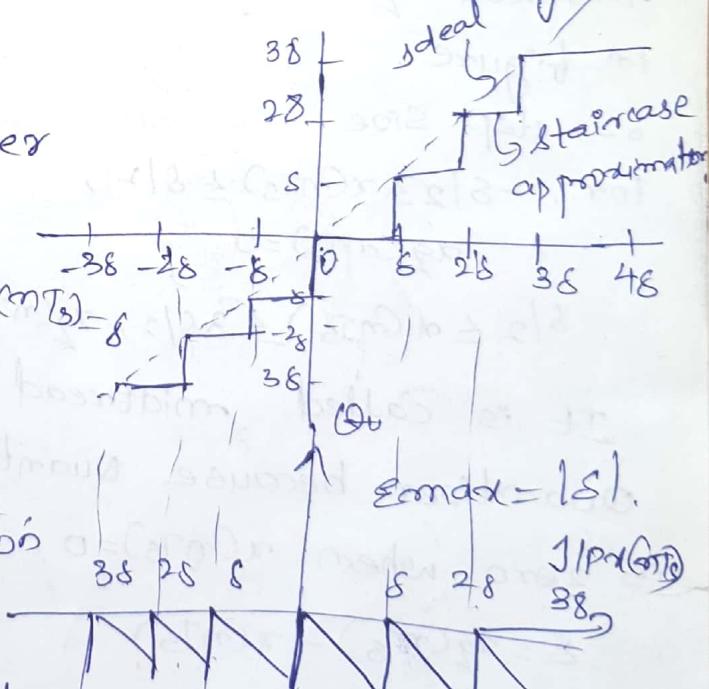
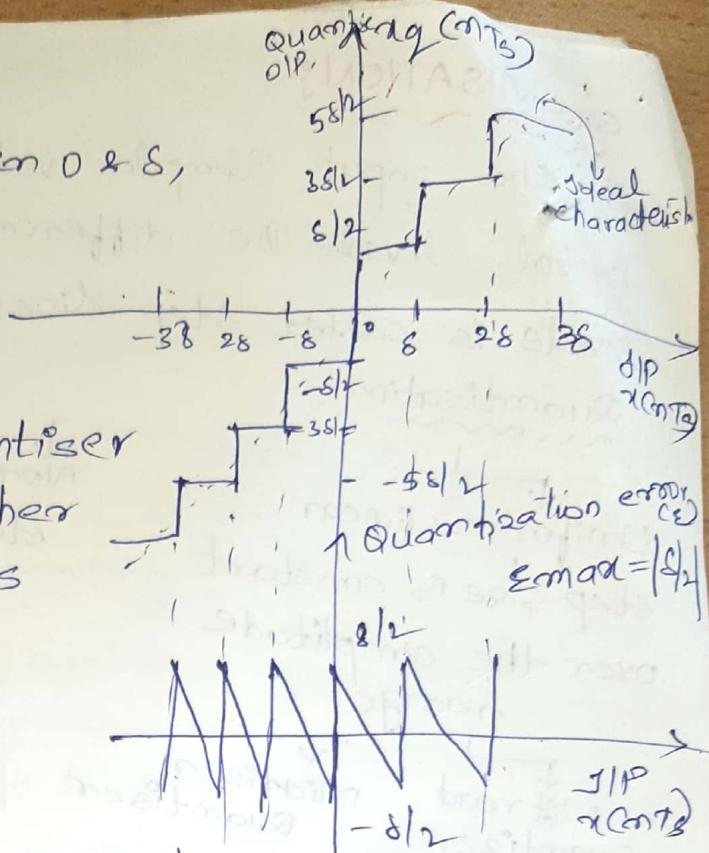
error is $\epsilon_{max} = \delta$.

The difference between staircase

and dotted line is called Quantisation error

- sation error. Hence, Quantisation error is

more in Biased Quantiser when Compared To
midtread & Midriser Quantiser.



Quantisation Noise

(3)

The noise which occur in Quantisation process
is called Quantisation Noise.

Noise in PCM

Quantisation error (ϵ) = $x_q(\text{nts}) - x(\text{nts})$

Consider an IIP signal $x(\text{nts})$ be of continuous amplitude which is in the range $-x_{\text{max}} \text{ to } x_{\text{max}}$.

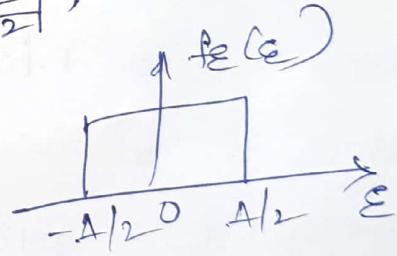
The total amplitude range =

$x_{\text{max}} - (-x_{\text{max}}) = 2x_{\text{max}}$
If this amplitude range is divided into 2 level quantiser, then the step size will be considered as $\Delta = \frac{2x_{\text{max}}}{2} = x_{\text{max}}$, if $x_{\text{max}} = 1$

$$\Delta = \frac{2}{2}$$

Quantisation error $E_{\text{max}} = \left| \frac{1}{2} \right|$,

$$f_{\epsilon}(\epsilon) = \begin{cases} 0, & \epsilon \in [-\Delta/2, \Delta/2] \\ \frac{1}{\Delta}, & -\Delta/2 < \epsilon < \Delta/2 \\ 0, & \epsilon > \Delta/2 \end{cases}$$



Signal - Noise Power = $\frac{\text{Noise}}{R}$

$$\sqrt{\text{noise}} = \bar{x} = E[x^2] = \int_{-\infty}^{\infty} x^2 \cdot f_x(x) dx = \int_{-\Delta/2}^{\Delta/2} x^2 \cdot \frac{1}{\Delta} dx = \frac{\Delta^2}{12}$$

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

Normalised S/N = $\frac{\text{Signal Power}}{\text{Noise Power}}$

$$q = 2^V, \Delta = \frac{2x_{\text{max}}}{q} = \frac{2x_{\text{max}}}{2^V} = 2^{V-1}$$

$$S/N = \frac{3P}{x_{\text{max}}^2} \cdot 2^V$$

Noise fm 8

Assume that slope overload distortion (from noise) occurs when $A_m \leq \frac{\Delta}{2\pi f_m T_s}$

If max. signal amplitude is there

$$A_m = \frac{\Delta}{2\pi f_m T_s}$$

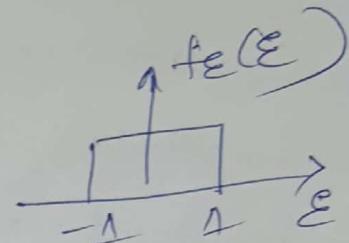
$$P = \frac{V_{rms}^2}{R} = \frac{(A_m)^2}{R} = \frac{A_m^2}{2}$$

$$P = \frac{\Delta^2}{4\pi^2 f_m^2 T_s^2 C_2}$$

$$P_s = \frac{\Delta^2}{8\pi^2 f_m T_s^2}$$

$$\text{Noise Power } P_{noise} = \int_{-\Delta/2}^{\Delta/2} \epsilon^2 \cdot \frac{1}{2\Delta} d\epsilon$$

$$P_{noise} = \Delta^2 / 3$$



This noise power is uniformly distributed over $-f_s$ to $+f_s$. At the OIP of 8 modulator receiver there is low pass reconstruction filter whose cutoff frequency is ω_{c2} . This cutoff frequency is equal to highest sign of frequency.

$$\begin{aligned} \text{OIP noise power} &= (\omega_{fs}) \text{ noise power} \\ &= (\omega_{fs}) (\Delta^2 / 3) = \omega_{fs} \Delta^2 / 3 \end{aligned}$$

$$S/N = \frac{P_{signal}}{P_{noise}} = \frac{\Delta^2}{8\pi^2 f_m^2 T_s^2} \times \frac{3}{\omega_{fs} \Delta^2}$$

$$S/N = \frac{3}{8\pi^2 f_m^2 \omega_{fs}^3}$$

Companding in PCM

④

The process in which used to achieve the non uniform quantisation is called. It is used mainly in order to improve SNR ratio of weak signals. Companding is the process of both compression & expansion. Compression is the process of amplifying the weak signal and attenuation of strong signals. Compression at the transmission and expansion at receiver is called Companding.

μ-law Companding

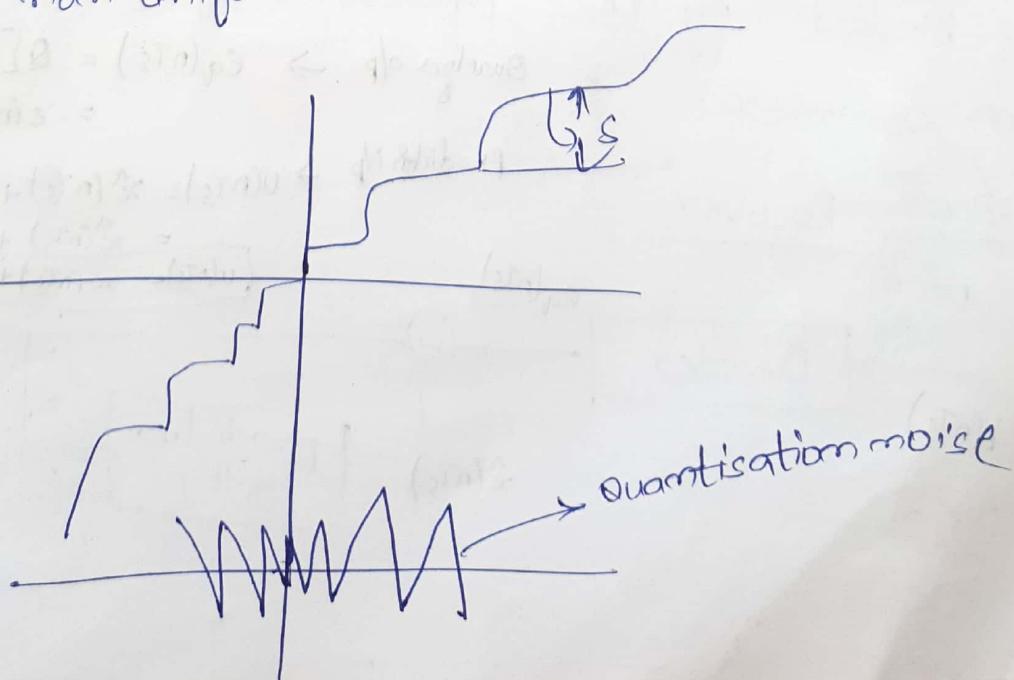
$$z(x) = \text{sgn}(x) \ln \frac{(1+\mu|x|)}{\ln(1+\mu)}, |x| \leq 1$$

A-law Companding

$$z(x) = \frac{A|x|}{1 + \ln A} \quad \text{for } 0 \leq |x| \leq \frac{1}{A}$$

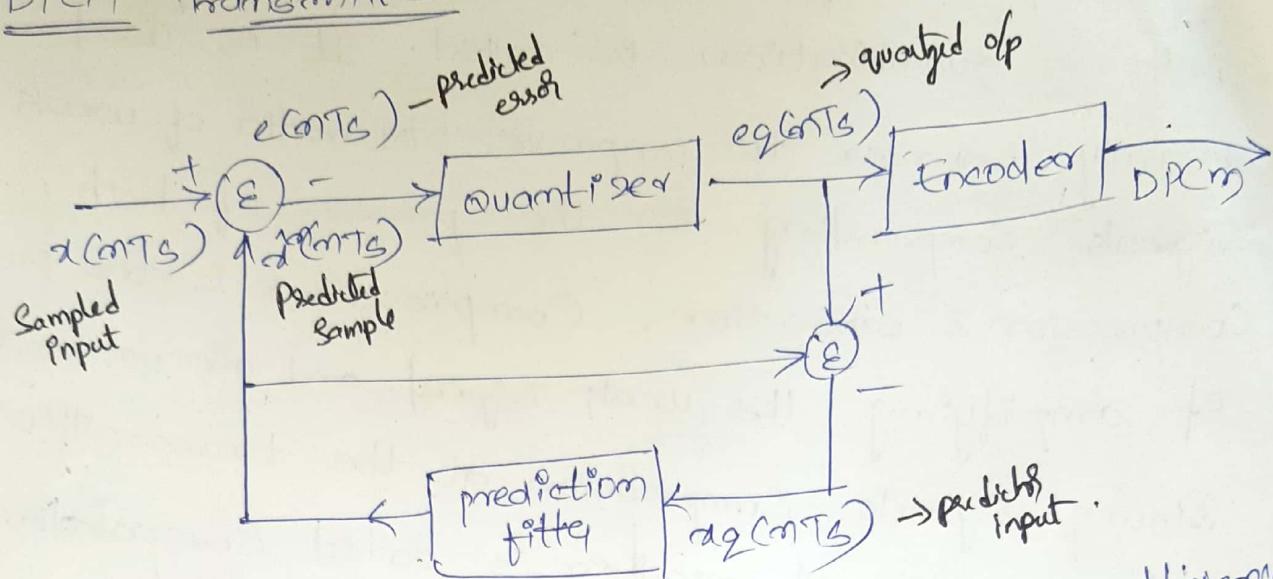
NON UNIFORM QUANTISATION

If step size is not constant, then it is called as non uniform quantisation



DPCM Differential Pulse Code Modulation

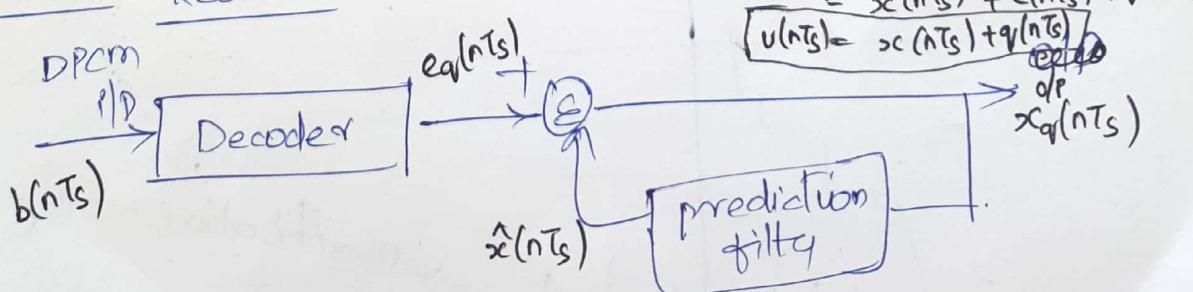
DPCM Transmitter



If the redundancy is reduced in PCM, efficiency will be increased. This will be done by DPCM.

DPCM works on the principle of prediction. The value of present sample is predicted from past sample signal which is denoted by $\hat{x}(nT_s)$. The comparator finds difference between present & previous sample. If both are same, that sample will be blocked by the prediction filter and later stored in summer.

DPCM Receiver



$$\text{Quantizer o/p} \rightarrow e_q(nT_s) = Q[e(nT_s)] \\ = e(nT_s) + q(nT_s)$$

$$\text{Predictor o/p} \rightarrow u(nT_s) = \hat{x}(nT_s) + e_q(nT_s) \\ = \hat{x}(nT_s) + e(nT_s) + q(nT_s)$$

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

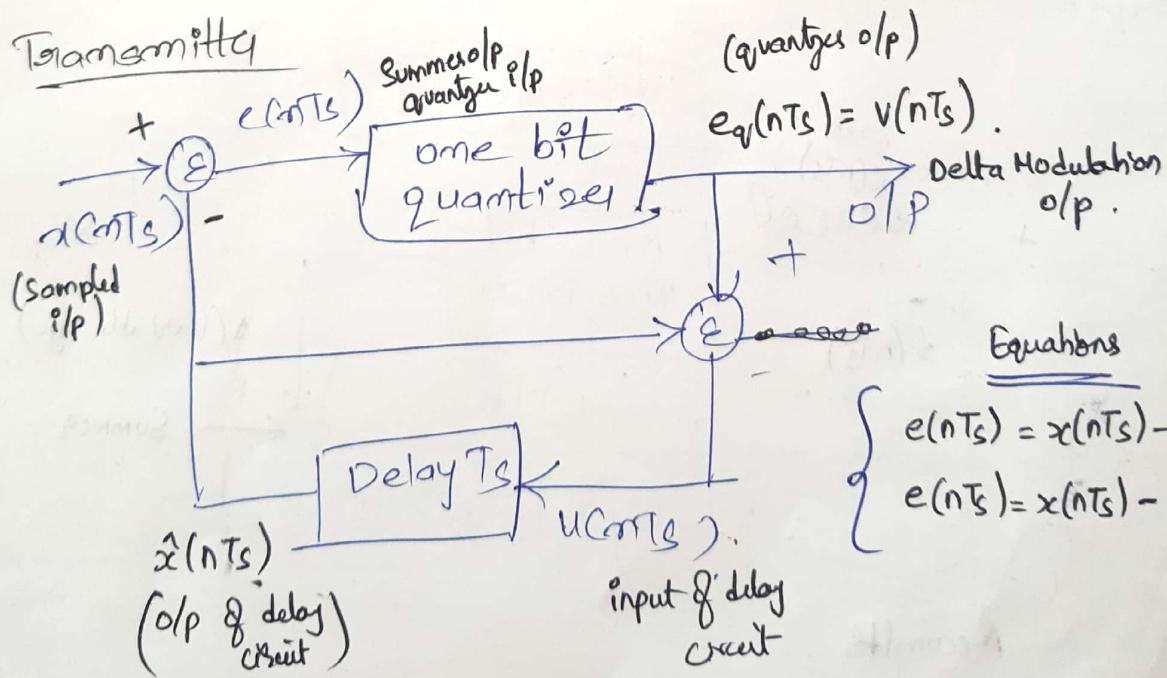
Q/P o/p

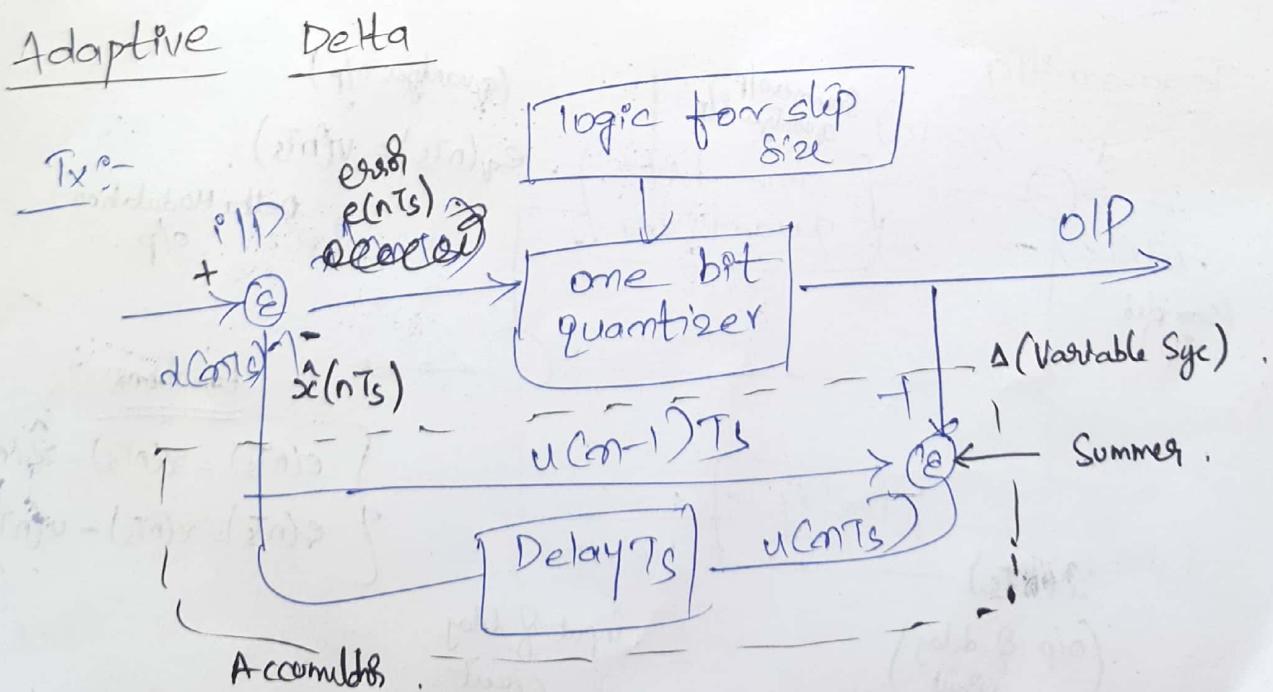
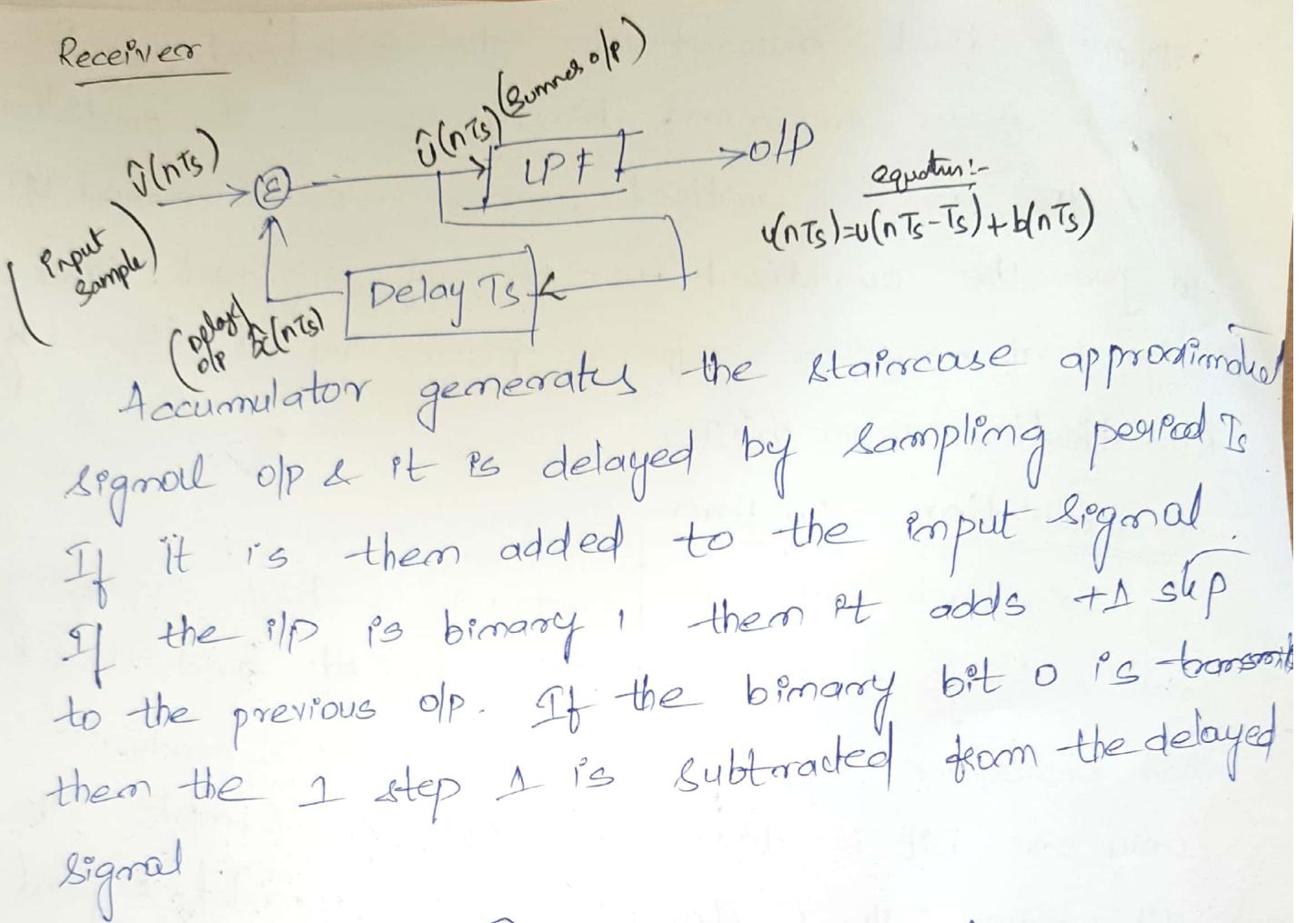
$$x_q(nT_s)$$

Decoder first reconstructs the quantised error signal from incoming binary signal. The prediction filter are Quantised errors are summed up to give the quantised version of original. Thus signal at receiver refers from actual signal by Quantisation error $e(nT_s)$

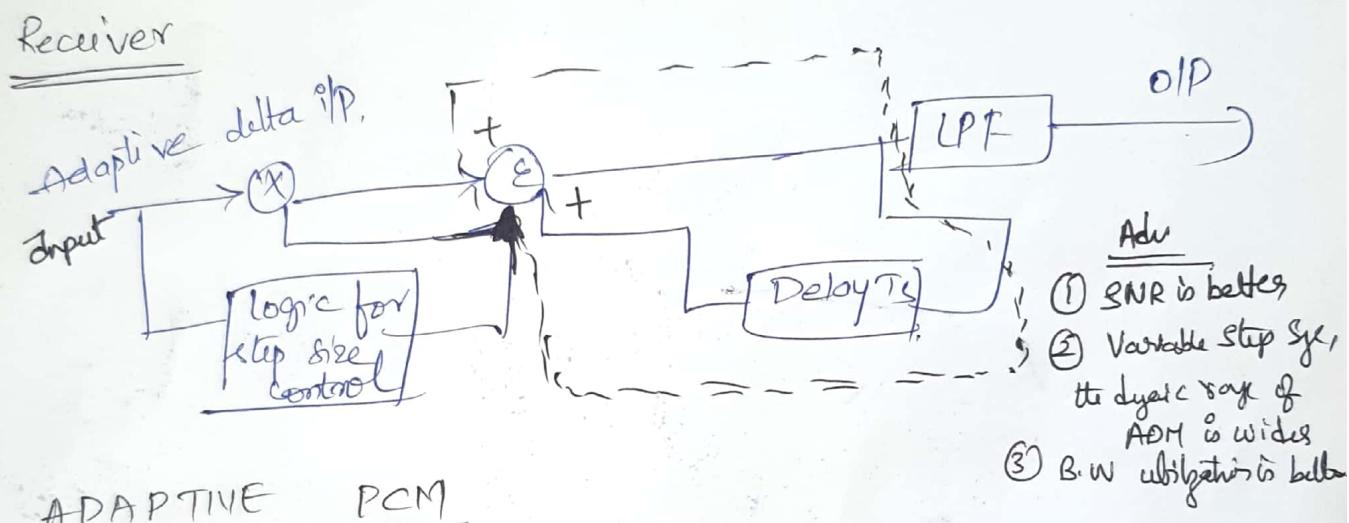
S-modulation system

Drawbacks of PCM system is high signalling rate & large transmission bandwidth and these can overcome by S modulation. In S modulation only one bit is transmitted per one sample. The difference is a staircase approximated signal confined to two levels +S and -S. If the step size is increased by +S, bit '1' is transmitted, otherwise bit '0' transmitted.

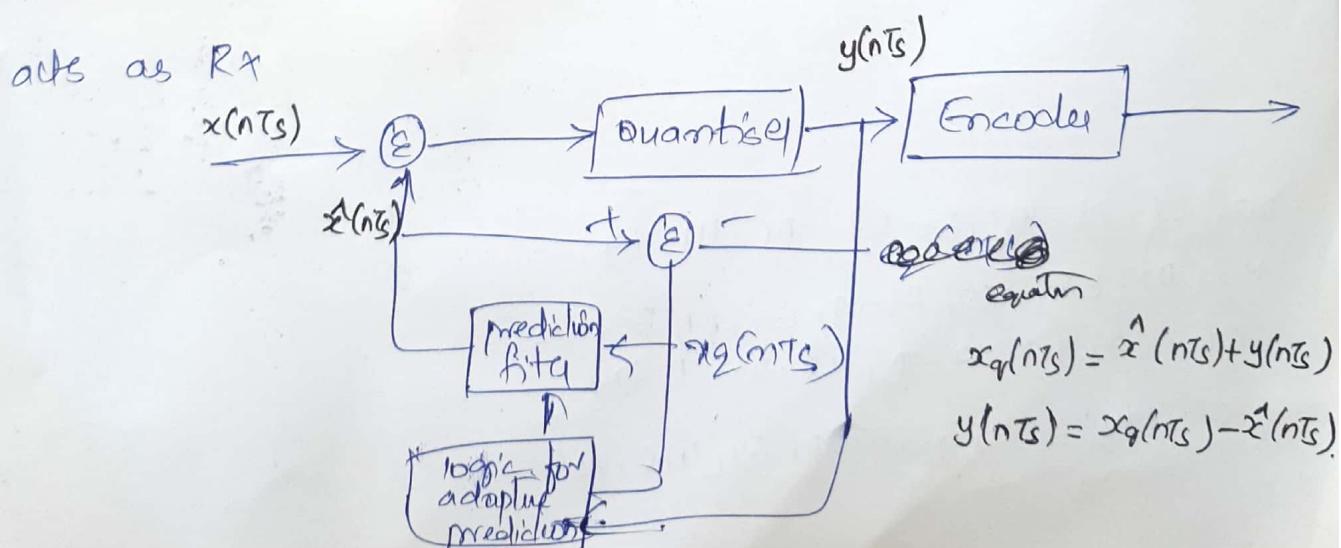




In order to overcome the disadvantages of Δ modulation, we go for adaptive delta modulation. Logic for step size control is added in figure. Step size is increased or decreased depending on certain rule. For eg, if one bit quantized O/P is high then step size will be doubled for next sample. If one bit quantized O/P is low, then step size will be reduced by one step.



It is a combination of Adaptive PCM, DPCM & Δ modulation. In forward direction, the above circuit acts as Tx & reverse direction it acts as Rx.



Step size depends upon amplitude level & spectrum of speech signal.

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