

### UNIT - III - (Information Theory)

#### SOURCE CODES, LINE CODES & ERROR CONTROL CODES

##### Primary Communication:

→ Carry information-bearing baseband signal from one place to another through a communication channel.

→ The Performance of communication system is measured in terms of Probability of error  $p_e$  i.e. if no error  $p_e$  will be zero.

##### Amount of information:

Let us consider the communication system which transmits messages  $m_1, m_2, \dots$  with probabilities  $P_1, P_2, \dots$ . Therefore, the amount of information  $I_{x_i}$  is related to the logarithm on the inverse of the Probability of occurrence of an event  $P(x_i)$ .

$$I_{x_i} = \log \frac{1}{P(x_i)}$$

$$x_i = k$$

i.e. 
$$I_k = \log \frac{1}{P_k}$$

Units of information

- (i) base '2'  $\rightarrow$  unit is bit,
- (ii) base '10'  $\rightarrow$  " " decit
- (iii) base 'e'  $\rightarrow$  " " nat

Default base is '2',

$$I_k = \log_2 \left( \frac{1}{p_k} \right)$$

for eg:

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

$$I_k = \log_2(2)$$

$$I_k = 1 \text{ bit}$$

Properties of information:

④ Property

$$(i) I(x_j) = 0 \text{ for } p(x_j) = 1$$

$$I(x_j) = \infty \text{ for } p(x_j) = 0$$

This means no information gained.

(ii)  ~~$I(x_j)$~~  Non-negative quantity i.e

$$I(x_j) \geq 0 \text{ for } 0 \leq p(x_j) \leq 1$$

This means no loss of information.

$$(iii) I(x_j) > I(y_k) \text{ for } p(x_j) < p(y_k)$$

This ~~fact~~ means more information gain.

$$(iv) I(x_j, y_k) = I(x_j) + I(y_k)$$

if  $x_j$  &  $y_k$  are statistically independent

### Entropy (or) Average information

It is defined as the "source which produces average information per message in a particular interval". It is also called as Comentropy.

Let  $m_1, m_2, \dots, m_k$  be  $k$  different messages with corresponding probabilities  $p_1, p_2, \dots, p_k$ .

Let us assume that ' $L$ ' sequence of messages have been generated for a long time interval with  $L \gg k$ .

Then, the number of messages

$$m_1 = p_1 L \longrightarrow (1)$$

The amount of information in message  $M_1$  is

$$I_1 = \log_2 \left( \frac{1}{p_1} \right) \longrightarrow (2)$$

The total amount of information due to message  $m_1$  is,  $I_L = I \times m$

iii) the total amount of information due to message  $M_2$  is

$$I_{t2} = p_2 \cdot \log_2 \left( \frac{1}{p_2} \right) \longrightarrow \textcircled{4}$$

The total amount of information due to sequence of  $L$  messages is given as

$$I_t = I_{t1} + I_{t2} + \dots + I_{tK}$$

$$= p_1 \cdot \log_2 \left( \frac{1}{p_1} \right) + p_2 \cdot \log_2 \left( \frac{1}{p_2} \right) + \dots$$

$$p_K \cdot \log_2 \left( \frac{1}{p_K} \right)$$

The average information per message is called entropy & it is represented by  $H$  (or)  $H(s)$ .

$$H(s) = \frac{I_t}{L} = p_1 \log_2 \left( \frac{1}{p_1} \right) + p_2 \log_2 \left( \frac{1}{p_2} \right) + \dots + p_K \log_2 \left( \frac{1}{p_K} \right)$$

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

Entropy is called as discrete memoryless source because each & every symbol emitted at any time are independent of the previous ones.

### Properties of Entropy:

The entropy of a discrete memoryless source is bounded as,

$$0 \leq H \leq \log_2 K$$

From this, we can define three properties of entropy

- (i)  $H = 0$  if  $p_k = 0$  (or)  $p_k = 1$ .
- (ii)  $H = \log_2 K$  when the symbols are equally likely for  $K$ -symbols i.e.  $p_k = 1/K$ .
- (iii) Maximum upper bound on entropy is

$$H_{\max} \leq \log_2 K$$

### Proof:

$$(i) \quad H = 0$$

when  $p_k = 0$

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

$$= p_1 \log_2 \left( \frac{1}{p_1} \right) + \dots + p_K \log_2 \left( \frac{1}{p_K} \right)$$

$$\boxed{H = 0}$$

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

$$\frac{\log_{10} 1}{\log_{10} 2}$$

(ii)  $H = \log_2 K$  when  $p_k = 1/K$ .

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

$$= p_1 \log_2 \left( \frac{1}{p_1} \right) + p_2 \log_2 \left( \frac{1}{p_2} \right) + \dots + p_K \log_2 \left( \frac{1}{p_K} \right)$$

$$= \frac{1}{K} \log_2 K + \frac{1}{K} \log_2 K + \dots + \frac{1}{K} \log_2 K$$

Adding the <sup>above</sup> ~~adding~~ equation we get  $K$  number of terms

$$H = \frac{K}{K} \log_2 K$$

(iii) Maximum upper bound on entropy is

$$H_{\max} \leq \log_2 K.$$

Proof:

This can be proved by using natural logarithms

$$\ln x \leq (x-1) \quad \text{for } x \geq 0.$$

Let us consider two probability distributions

$$\{p_1, p_2, \dots, p_K\} \text{ \& \& } \{q_1, q_2, \dots, q_K\}.$$

W.K.T

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

It can be written as

$$\sum_{k=1}^K p_k \log_2 \left( \frac{q_k}{p_k} \right) = \sum_{k=1}^K p_k \frac{\log_{10} \left( \frac{q_k}{p_k} \right)}{\log_{10} 2}$$

Multiply and divide by  $\log_{10} e$  in R.H.S

$$\sum_{k=1}^K p_k \log_2 \left( \frac{q_k}{p_k} \right) = \sum_{k=1}^K p_k \frac{\log_{10} e}{\log_{10} e} \frac{\log_{10} \left( \frac{q_k}{p_k} \right)}{\log_{10} 2}$$

$$= \sum_{k=1}^K p_k \frac{\log_{10} e}{\log_{10} 2} \cdot \frac{\log_{10} \left( \frac{q_k}{p_k} \right)}{\log_{10} e}$$

$$= \sum_{k=1}^K p_k \log_2 e \cdot \log_e \left( \frac{q_k}{p_k} \right)$$

from natural logarithm

$$\log_e \left( \frac{q_k}{p_k} \right) = \ln \left( \frac{q_k}{p_k} \right)$$

$$\sum_{k=1}^K p_k \log_2 \left( \frac{q_k}{p_k} \right) = \sum_{k=1}^K p_k \log_2 e \cdot \ln \left( \frac{q_k}{p_k} \right)$$

$$= \log_2 e \sum_{k=1}^K p_k \ln \left( \frac{q_k}{p_k} \right)$$

$$\ln\left(\frac{q_k}{p_k}\right) \leq \left(\frac{q_k}{p_k} - 1\right).$$

$$\sum_{k=1}^K p_k \log_2\left(\frac{q_k}{p_k}\right) \leq \log_2 e \sum_{k=1}^K p_k \left(\frac{q_k}{p_k} - 1\right)$$

$$\leq \log_2 e \sum_{k=1}^K p_k \left(\frac{q_k - p_k}{p_k}\right)$$

$$\leq \log_2 e \sum_{k=1}^K (q_k - p_k)$$

$$\leq \log_2 e \left( \sum_{k=1}^K q_k - \sum_{k=1}^K p_k \right)$$

$$\sum_{k=1}^K p_k \log_2\left(\frac{q_k}{p_k}\right) \leq \log_2 e (1 - 1)$$

$$\sum_{k=1}^K p_k \log_2\left(\frac{q_k}{p_k}\right) \leq 0$$

$$\sum_{k=1}^K p_k \left[ \log_2 q_k + \log_2 \left(\frac{1}{p_k}\right) \right] \leq 0$$

$$\sum_{k=1}^K p_k \log_2(q_k) + \sum_{k=1}^K p_k \log_2\left(\frac{1}{p_k}\right) \leq 0$$

$$\sum_{k=1}^K p_k \log_2\left(\frac{1}{p_k}\right) \leq - \sum_{k=1}^K p_k \log_2 q_k$$

$$\sum_{k=1}^K p_k \log_2 \left( \frac{1}{p_k} \right) \leq \sum_{k=1}^K p_k \log_2 \left( \frac{1}{q_k} \right).$$

Let us assume all symbols in the alphabet are equally likely. Suppose we next put

$$\text{Then } q_k = \frac{1}{K}$$

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

$$\leq \log_2 K \sum_{k=1}^K p_k$$

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

The L.H.S of the above eqn is called entropy.

$$H \leq \log_2 K$$

Maximum value of entropy is

$$H_{\max} \leq \log_2 K$$

① Find the amount of info & its entropy when a source emits 5 symbols

Amount of info due to symbol ' $s_i$ ' is

$$(i) I_1 = \log_2 \left( \frac{1}{p_1} \right) = \log_2 \left( \frac{1}{0.2} \right)$$

$$= \log_2 \left( \frac{5}{1} \right)$$

$$= \frac{\log_{10} \left( \frac{5}{1} \right)}{\log_{10} 2}$$

$$= 2.3219 \text{ bits}$$

$$(ii) I_2 = \log_2 \left( \frac{1}{p_2} \right) = \log_2 \left( \frac{1}{0.3} \right)$$

$$= \frac{\log_{10} \left( \frac{1}{0.3} \right)}{\log_{10} 2}$$

$$= 1.9367 \text{ bits}$$

$$(iii) I_3 = 2.3219 \text{ bits}$$

$$(iv) I_4 = 2.3219 \text{ bits}$$

$$(v) I_5 = 3.3219 \text{ bits}$$

Entropy

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

$$= p_1 \log_2 \left( \frac{1}{p_1} \right) + p_2 \log_2 \left( \frac{1}{p_2} \right) +$$

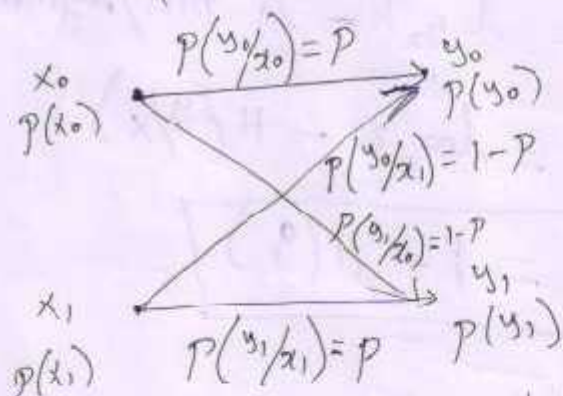
$$= 0.2 \times 2.3219 + 0.3 \times 1.7369 + 0.2 \times 2.3219 + 0.2 \times 2.3219 + 0.1 \times 3.3219$$

$$= 0.4644 + 0.5211 + 0.4644 + 0.4644 + 0.3322$$

$$H = 2.246 \text{ bits/symbol.}$$

BSC (Binary Symmetric channel)  
Binary Communication channel is said to be symmetric if

$$P(y_0/x_0) = P(y_1/x_1) = P$$



It can be represented using the matrix

$$P(y) = P(x) P(y/x)$$

$$\begin{bmatrix} p(y_0) \\ p(y_1) \end{bmatrix} = \begin{bmatrix} p(x_0) & p(x_1) \end{bmatrix} \begin{bmatrix} P(y_0/x_0) & P(y_1/x_0) \\ P(y_0/x_1) & P(y_1/x_1) \end{bmatrix}$$

$$\begin{bmatrix} P(y_0) \\ P(y_1) \end{bmatrix} = \begin{bmatrix} P(x_0) & P(x_1) \end{bmatrix} \begin{bmatrix} P & 1-P \\ 1-P & P \end{bmatrix}$$

In Binary Symmetric channel, the value of  $m=n=2$  (number of rows & columns),

$\therefore k=2$  &

The channel matrix  $\mathcal{Q} = P(y/x)$

$$= \begin{bmatrix} P & 1-P \\ 1-P & P \end{bmatrix}$$

The channel capacity for a binary symmetric channel is,

$$C = \log_2 k - A \text{ bits/symbol}$$

$$= \log_2 2 - H(y/x)$$

$$C = 1 - H(q)$$

① A BSC is shown as



- (i) find the channel matrix  
 (ii) find  $P(y_0)$  &  $P(y_1)$  if  $P(x_0) = 0.3$ ,  $P(x_1) = 0.7$ .

Ans

$$P(y_0/x_0) = 0.6$$

$$P(y_0/x_1) = 0.4$$

$$P(y_1/x_0) = 0.4$$

$$P(y_1/x_1) = 0.6$$

(i) channel matrix  $P(y/x) = \begin{bmatrix} P(y_0/x_0) & P(y_1/x_0) \\ P(y_0/x_1) & P(y_1/x_1) \end{bmatrix}$

$$= \begin{bmatrix} 0.6 & 0.4 \\ 0.4 & 0.6 \end{bmatrix}$$

$$= 0.56 - 0.16$$

$$P(y/x) = 0.2$$

(ii)  $P(y) = P(x) P(y/x)$

$$\begin{bmatrix} P(y_0) \\ P(y_1) \end{bmatrix} = \begin{bmatrix} P(x_0) & P(x_1) \end{bmatrix} P(y/x)$$

$$= \begin{bmatrix} 0.3 & 0.7 \end{bmatrix} \begin{bmatrix} 0.6 & 0.4 \\ 0.4 & 0.6 \end{bmatrix}$$

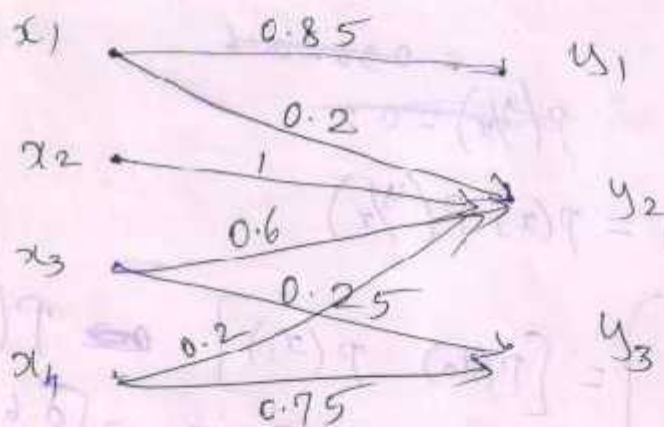
$$= \begin{bmatrix} 0.46 & 0.54 \\ 0.46 & 0.54 \end{bmatrix}$$

$$P(y) = \begin{bmatrix} P(y_0) & P(y_1) \end{bmatrix} \quad P(y_0) = 0.46$$

② Draw the Binary asymmetric channel diagram for

$$P(y/x) = \begin{bmatrix} \cancel{0.85} & 0.2 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0.6 & 0.25 & 0 \\ 0 & 0.2 & 0.75 & 0 \end{bmatrix}$$

$$P(y/x) = \begin{bmatrix} P(y_1/x_1) & P(y_2/x_1) & P(y_3/x_1) & P(y_4/x_1) \\ P(y_1/x_2) & P(y_2/x_2) & P(y_3/x_2) & P(y_4/x_2) \\ P(y_1/x_3) & P(y_2/x_3) & P(y_3/x_3) & P(y_4/x_3) \\ P(y_1/x_4) & P(y_2/x_4) & P(y_3/x_4) & P(y_4/x_4) \end{bmatrix}$$

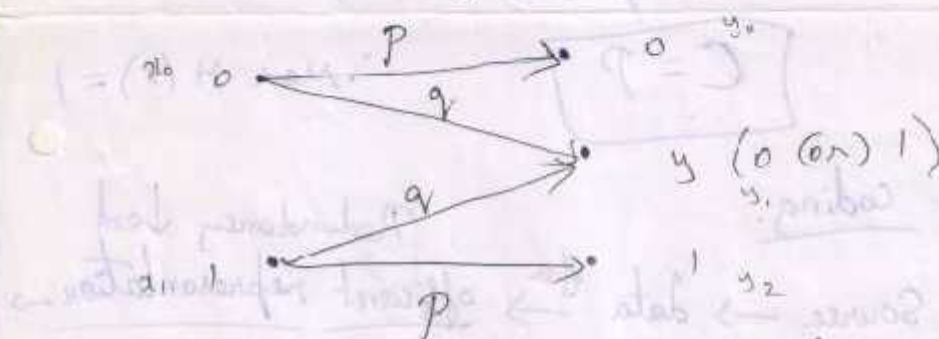


## Binary Erasure channel

It has two inputs 0, 1 & three outputs 0, y, 1 i.e. we transmitted 0, 1 & we received 0, y, 1.

y indicates due to noise, that can't be Predicted, whether it is 0 & 1. Alternatively y indicates that the o/p is erased.

### channel Diagram



$$\text{channel matrix} = \begin{bmatrix} P(y_0/x_0) & P(y_1/x_0) & P(y_2/x_0) \\ P(y_0/x_1) & P(y_1/x_1) & P(y_2/x_1) \end{bmatrix}$$

$$\text{channel matrix} = \begin{bmatrix} P & q & 0 \\ 0 & q & P \end{bmatrix}$$

channel capacity  $C = \text{Max } I(x; y)$

$$I(x; y) = H(x) - H(x|y)$$

$$= H(x) - H(x)(1-p)$$

$$= H(x) [1 - 1 + p]$$

$$= p H(x)$$

$$C = \text{Max } p H(x)$$

$$= p \text{ Max } H(x)$$

$$\boxed{C = p}$$

$$\therefore \text{Max } H(x) = 1$$

Source coding:

Redundancy ↓

Source → data → efficient representation →

Source coding is used.

- (i) Shannon Fano coding → used to encode the messages depending upon their probabilities
- (ii) Huffman coding

Shannon Fano Coding:Procedure:

Step 1: List the source symbols in order of decreasing probabilities

Step 2: Partition the set into two sets that are as close to equiprobable as possible and assign 0 to the upper set and assign 1 to the lower set.

Step 3: Continue this process each time partitioning the sets with as nearly probabilities as possible until further partitioning is not possible.

- ① A discrete memoryless source has 5 symbols  $S_1, S_2, S_3, S_4, S_5$  with probabilities 0.4, 0.2, 0.1, 0.2, 0.1 respectively. Construct a Shannon Fano coding and calculate its efficiency.

Given

Symbols	Probabilities
$S_1$	0.4
$S_2$	0.2
$S_3$	0.1
$S_4$	0.2
$S_5$	0.1

Symbols	Probabilities	Stage 1	Stage 2	CW
$S_1$	0.4	0.4	0	0
$S_2$	0.2	0.2	0	10
$S_4$	0.2	0.2	0	110
$S_3$	0.1	0.1	0	1110
$S_5$	0.1	0.1	1	1111

CW	Length
0	1
10	2
110	3
1110	4
1111	4

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

$$\begin{aligned} \bar{L} &= \sum_{k=1}^K P_k L_k = \sum_{k=1}^5 P_k L_k \\ &= P_1 L_1 + P_2 L_2 + P_3 L_3 + P_4 L_4 + P_5 L_5 \\ &= (0.4)(1) + (0.2)(2) + (0.2)(3) + (0.1)(4) + (0.1)(4) \\ &= 0.4 + 0.4 + 0.6 + 0.4 + 0.4 \\ &= 2.2 \text{ bits/symbol} \end{aligned}$$

$$\begin{aligned} H &= \sum_{k=1}^K P_k \log_2 \left( \frac{1}{P_k} \right) \\ &= P_1 \log_2 \left( \frac{1}{P_1} \right) + P_2 \log_2 \left( \frac{1}{P_2} \right) + P_3 \log_2 \left( \frac{1}{P_3} \right) + P_4 \log_2 \left( \frac{1}{P_4} \right) + P_5 \log_2 \left( \frac{1}{P_5} \right) \\ &= (0.4) \log_2 \left( \frac{1}{0.4} \right) + (0.2) \log_2 \left( \frac{1}{0.2} \right) + (0.2) \log_2 \left( \frac{1}{0.2} \right) + (0.1) \log_2 \left( \frac{1}{0.1} \right) + (0.1) \log_2 \left( \frac{1}{0.1} \right) \\ &= (0.4) \frac{\log_{10} \left( \frac{1}{0.4} \right)}{\log_{10} 2} + (0.2) \frac{\log_{10} \left( \frac{1}{0.2} \right)}{\log_{10} 2} + (0.2) \frac{\log_{10} \left( \frac{1}{0.2} \right)}{\log_{10} 2} + \\ &\quad (0.1) \frac{\log_{10} \left( \frac{1}{0.1} \right)}{\log_{10} 2} + (0.1) \frac{\log_{10} \left( \frac{1}{0.1} \right)}{\log_{10} 2} \end{aligned}$$

$$= 0.5287 + 0.4644 + 0.4644 + 0.3322 + 0.3322$$

$$H = 2.122 \text{ bits/symbols}$$

$$\eta = \frac{H}{L} = \frac{2.122}{2.2} = 0.9645$$

$$\boxed{\eta = 96.45\%}$$

- ① A discrete memoryless source has emits the following 6 messages  $m_1, m_2, m_3, m_4, m_5, m_6$  with Probabilities 0.3, 0.25, 0.05, 0.12, 0.08, 0.2 respectively. Compute the Shannon fano code & also calculate its  $\eta$ .

Symbol	Probabilities	stage 1	stage 2	stage 3	Stage 4
$m_1$	0.3	0.55 <div> <div>0</div> <div>0</div> </div>	0.3 <div>0</div>	0.25 <div>1</div>	
$m_2$	0.25				
$m_6$	0.2	0.45 <div>1</div> <div>1</div> <div>1</div> <div>1</div>	0.2 <div>0</div>	0.12 <div>0</div>	0.08 <div>0</div>
$m_4$	0.12				
$m_5$	0.08				
$m_3$	0.05				

CW	Length
00	2
01	2
10	2
110	3
1110	4
1111	4

$$\bar{L} = \sum_{k=1}^6 P_k L_k$$

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

$$= (0.3)(2) + (0.25)(2) + (0.05)(2) + (0.12)(3) + (0.08)(4) + (0.2)(4)$$

$$\begin{aligned}
 H &= \sum_{k=1}^6 p_k \log_2 \left( \frac{1}{p_k} \right) \\
 &= p_1 \log_2 \left( \frac{1}{p_1} \right) + p_2 \log_2 \left( \frac{1}{p_2} \right) + p_3 \log_2 \left( \frac{1}{p_3} \right) + \\
 &\quad p_4 \log_2 \left( \frac{1}{p_4} \right) + p_5 \log_2 \left( \frac{1}{p_5} \right) + p_6 \log_2 \left( \frac{1}{p_6} \right) \\
 &= 0.3 \log_2 \left( \frac{1}{0.3} \right) + (0.25) \log_2 \left( \frac{1}{0.25} \right) + (0.2) \log_2 \left( \frac{1}{0.2} \right) \\
 &\quad + (0.12) \log_2 \left( \frac{1}{0.12} \right) + (0.08) \log_2 \left( \frac{1}{0.08} \right) + (0.05) \log_2 \left( \frac{1}{0.05} \right) \\
 &= 0.3 \frac{\log_{10} \left( \frac{1}{0.3} \right)}{\log_{10} 2} + (0.25) \frac{\log_{10} \left( \frac{1}{0.25} \right)}{\log_{10} 2} + (0.2) \frac{\log_{10} \left( \frac{1}{0.2} \right)}{\log_{10} 2} \\
 &\quad + (0.12) \frac{\log_{10} \left( \frac{1}{0.12} \right)}{\log_{10} 2} + (0.08) \frac{\log_{10} \left( \frac{1}{0.08} \right)}{\log_{10} 2} + \\
 &\quad (0.05) \frac{\log_{10} \left( \frac{1}{0.05} \right)}{\log_{10} 2} \\
 &= (0.3)(1.7369) + (0.25)(2) + (0.2)(2.3219) + \\
 &\quad (0.12)(3.0588) + (0.08)(3.6438) + \\
 &\quad (0.05)(4.3219) \\
 &= 0.52107 + 0.5 + 0.4644 + 0.36696 + 0.2915 + \\
 &\quad 0.2161
 \end{aligned}$$

$$\begin{aligned}
 & \frac{\log_{10} 2}{\log_{10} 2} + \frac{\log_{10} 2}{\log_{10} 2} + \frac{(0.05) \log_{10} \left(\frac{1}{0.05}\right)}{\log_{10} 2} \\
 & = (0.3)(1.7369) + (0.25)(2) + (0.2)(2.3219) + \\
 & \quad (0.12)(3.0588) + (0.08)(3.6438) + \\
 & \quad (0.05)(4.3219) \\
 & = 0.52107 + 0.5 + 0.4644 + 0.36696 + 0.2915 + \\
 & \quad 0.2161 \\
 & H = 2.36003 \text{ inform bits/symbol} \\
 & \eta = \frac{H}{L} = \frac{2.36003}{2.38} = 0.9916 \\
 & \boxed{\eta = 99.1\%}
 \end{aligned}$$

## Huffman Coding:

→ also called as Minimum redundancy code (MRC)

Optimum code

### Procedure:

- Step 1: The Messages are arranged in an order of decreasing Probabilities.
- Step 2: The two messages of lowest Probabilities are assigned binary '0' & '1'.
- Step 3: The two lowest Probabilities in stage I are added & the sum is placed in stage II, such that Probabilities are in descending order.
- 4: Now last two Probabilities are assigned 0 & 1 & they are added. The sum of last two Probabilities placed in stage III such that Probabilities are in descending order. Again '0' & '1' is assigned to the last two Probabilities.
- 5: This Process continued till the last stage contains only two values. These two values are assigned digits 0 & 1 & no further repetition.

Example

- ① A discrete memoryless source has 5 symbols  $s_1, s_2, s_3, s_4, s_5$  with Probabilities 0.4, 0.2, 0.1, 0.2, 0.1 respectively. Construct a Huffman code & calculate its  $\frac{50}{n}$

Step 1: Rearrange the Probabilities in decreasing order

Symbols

Probabilities

$s_1$

0.4

$s_2$

0.2

$s_4$

0.2

$s_3$

0.1

$s_5$

0.1

Step 2:

Symbols

Stage 1

Stage 2

Stage 3

Stage 4

$s_1$

0.4 (00)

0.4 (00)

$s_2$

0.2 (10)

0.2 (01)

$s_4$

0.2 (11)

0.2 (10)

$s_3$

0.1 (010)

0.2 (11)

$s_5$

0.1 (011)

Step 3:

Symbols

Probabilities

Code word

Length (L)

$s_1$

0.4

00

2

$s_2$

0.2

10

2

$s_4$

0.2

11

2

$s_3$

0.1

010

3

$s_5$

0.1

011

3

Step 4To find  $\eta$ 

$$(i) \bar{L} = \sum_{k=1}^N p_k L_k$$

$$= \sum_{k=1}^5 p_k L_k$$

$$= p_1 L_1 + p_2 L_2 + p_3 L_3 + p_4 L_4 + p_5 L_5 + p_6 L_6$$

$$= (0.4)(2) + (0.2)(2) + (0.2)(2) + (0.1)(3) + (0.1)(3)$$

$$= 0.8 + 0.4 + 0.4 + 0.3 + 0.3$$

$$\bar{L} = 2.2 \text{ bits/symbol}$$

$$(ii) \text{Entropy } H(S) = \sum_{k=1}^6 p_k \log_2 \left( \frac{1}{p_k} \right)$$

$$= p_1 \log_2 \left( \frac{1}{p_1} \right) + p_2 \log_2 \left( \frac{1}{p_2} \right) +$$

$$p_3 \log_2 \left( \frac{1}{p_3} \right) + p_4 \log_2 \left( \frac{1}{p_4} \right) + p_5 \log_2 \left( \frac{1}{p_5} \right)$$

$$= (0.4) \log_2 \left( \frac{1}{0.4} \right) + (0.2) \log_2 \left( \frac{1}{0.2} \right) +$$

$$(0.2) \log_2 \left( \frac{1}{0.2} \right) + (0.1) \log_2 \left( \frac{1}{0.1} \right) +$$

$$(0.1) \log_2 \left( \frac{1}{0.1} \right)$$

$$= (0.4) \frac{\log_{10} \left( \frac{1}{0.4} \right)}{\log_{10} 2} + (0.2) \frac{\log_{10} \left( \frac{1}{0.2} \right)}{\log_{10} 2} + \dots$$

$$(0.1) \frac{\log_{10} \left( \frac{1}{0.1} \right)}{\log_{10} 2} + \dots$$

$$\eta = \frac{H}{L} = \frac{2.122}{2.2}$$

$$= 0.9645$$

$$\boxed{\eta = 96.45\%}$$

- ① A discrete memoryless source has an alphabet of 7 symbols whose probabilities of occurrence

Symbols	$x_0$	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	$x_6$
Probability	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{1}{16}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{1}{8}$

Compute

- (i) Huffman Code  
(ii) Calculate its  $\eta$ .

- ② A source emits 5 symbols with probabilities 0.15, 0.10, 0.05, 0.15, 0.55 respectively.

- Compute (i) Huffman Code  
(ii) Shannon Code  
(iii) Compare its average code word length.

① Symbols	Probability (stage 1)	stage 2	stage 3	stage 4	stage 5
$x_0$	$\frac{1}{4}^{(10)}$	$\frac{1}{4}^{(10)}$	$\frac{1}{4}^{(01)}$	$\frac{1}{4}^{(00)}$	$\frac{1}{2}^{(1)}$
$x_5$	$\frac{1}{4}^{(11)}$	$\frac{1}{4}^{(11)}$	$\frac{1}{4}^{(10)}$	$\frac{1}{4}^{(01)}$	$\frac{1}{4}^{(00)}$

Symbol	Probability	Code word	Length
$x_0$	$\frac{1}{4}$	10	2
$x_5$	$\frac{1}{4}$	11	2
$x_4$	$\frac{1}{8}$	001	3
$x_1$	$\frac{1}{8}$	010	3
$x_3$	$\frac{1}{8}$	011	3
$x_6$	$\frac{1}{16}$	0000	4
$x_2$	$\frac{1}{16}$	0001	4
$x_3$	$\frac{1}{16}$		

$$\bar{L} = \sum_{k=1}^7 p_k \cdot k$$

$$= \left(\frac{1}{4}\right)(2) + \left(\frac{1}{4}\right)(2) + \left(\frac{1}{8}\right)(3) + \left(\frac{1}{8}\right)(3) + \left(\frac{1}{8}\right)(3) +$$

$$\left(\frac{1}{16}\right)(4) + \left(\frac{1}{16}\right)(4)$$

$$= 0.5 + 0.5 + 0.375 + 0.375 + 0.375 + 0.25 + 0.25$$

$$\bar{L} = 2.625 \text{ bits/symbol}$$

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

$$= p_1 \log_2 \left( \frac{1}{p_1} \right) + \dots$$

stage 6      stage 7

$$= \left(\frac{1}{4}\right) \log_2(4) + \left(\frac{1}{4}\right) \log_2(4) +$$

$$\left(\frac{1}{8}\right) \log_2(8) + \left(\frac{1}{8}\right) \log_2(8) + \left(\frac{1}{8}\right) \log_2(8) +$$

$$+ \left(\frac{1}{16}\right) \log_2(16) + \left(\frac{1}{16}\right) \log_2(16)$$

~~$\eta = 100\%$~~   $\eta = 100\%$   $\eta = 1$

② Symbols      Probability (stage 1)      stage 2      stage 3      stage 4      stage 5

$x_5$	0.55 (0)	$\rightarrow$ 0.55 (0)	$\rightarrow$ 0.55 (0)	$\rightarrow$ 0.55 (0)	$\rightarrow$ 0.55 (0)
$x_1$	0.15 (100)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.30 (10)	$\rightarrow$ 0.45 (11)	$\rightarrow$ 0.45 (11)
$x_4$	0.15 (01)	$\rightarrow$ 0.15 (10)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.15 (11)
$x_2$	0.10 (110)	$\rightarrow$ 0.15 (101)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.15 (11)
$x_3$	0.05 (111)	$\rightarrow$ 0.15 (101)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.15 (11)	$\rightarrow$ 0.15 (11)

	Code word	Length
0.55	0	1
0.15	100	3
0.15	(0)	3
0.10	110	3
0.05	111	3

$$L = \sum_{k=1}^K p_k L_k = (0.55)(1) + (0.15)(3) + (0.15)(3) + (0.10)(3) + (0.05)(3)$$

$$= 0.55 + 0.45 + 0.45 + 0.3 + 0.15$$

$$= 1.9$$
  

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

$$= (0.55) \log_2 \left( \frac{1}{0.55} \right) + 0.15 \log_2 \left( \frac{1}{0.15} \right) +$$

$$(0.10) \log_2 \left( \frac{1}{0.10} \right) +$$

$$\begin{aligned}
 & \xrightarrow{K=1} \\
 & = (0.55) \log_2 \left( \frac{1}{0.55} \right) + 0.15 \log_2 \left( \frac{1}{0.15} \right) + \\
 & \quad (0.10) \log_2 \left( \frac{1}{0.10} \right) + \\
 & \quad (0.05) \log_2 \left( \frac{1}{0.05} \right) \\
 & = 0.4744 + 0.411 + 0.411 + 0.3322 + \\
 & \quad 0.2161 \\
 & = 1.8447 \quad \eta = \frac{H}{L} = \frac{1.8447}{1.9} = \frac{0.9708}{1} = 97.08\%
 \end{aligned}$$

## Shannon's Theorem (Noiseless Coding Theorem)

- (i) Shannon's first theorem (or) source coding theorem
- (ii) Shannon's second theorem (or) channel coding theorem
- (iii) Shannon's third theorem (or) Information capacity theorem (or) Shannon's Hartley's theorem

### (i) Shannon's first theorem:

Source coding theorem such as Shannon's zero coding, Huffman coding, Prefix coding is used to remove redundancy & to improve efficiency.

$$L \geq H$$

### (ii) Shannon's second theorem:

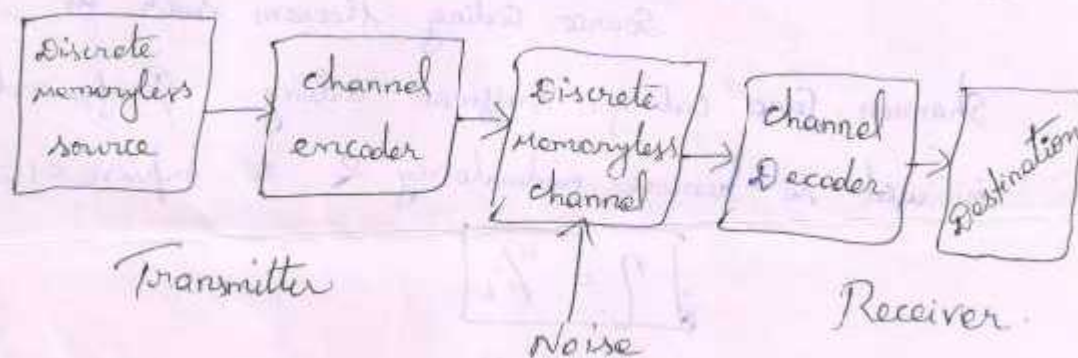
The presence of noise in the channel can be minimized using channel coding. For example: In wireless communication channel, the error probability may be  $10^{-1}$  (90% received correctly). But in many applications, this level of reliability is unacceptable. A probability

is often a necessary requirement.

To achieve this requirement, we are using channel coding.

The goal is to increase the resistance of digital communication systems to channel noise.

Process:



Approach:

To introduce redundancy in channel encoder so as to reconstruct the original source sequence as accurately as possible.

This theorem says that if

$R \leq C$ , it is possible to

transmit information without any error even

$$R = nH$$

The code rate is given as

$$\underline{n} = \frac{K}{N} \quad \text{always} < 1$$

$$\underline{\text{channel capacity}} / \underline{\text{unit}} = \frac{C}{T_c} \text{ bits/sec.}$$

Statement:

It is stated as two parts.

(i) Let DMS with alphabet  $S$  and entropy  $H(S)$  produce symbols for every  $T_s$  seconds.

Let DMC have capacity  $C$  & used for every  $T_c$  seconds.

$$\text{Then if } \frac{H(S)}{T_s} \leq \frac{C}{T_c}$$

There exists a coding scheme for which source output can be transmitted through the channel & can be reconstructed with small Probability of error.

Then the system is said to be signalling at a critical rate.

$$(ii) \text{ If } \frac{H(s)}{T_s} > \frac{C}{T_s}$$

No possibility of transmission & reconstruction with small probability of error.

(iii) Shannon's Third theorem:

→ channel in which noise is gaussian is known as Shannon's Hartley theorem. It is also called as information capacity theorem.

channel capacity of a white bandlimited gaussian channel is given by,

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

where  $B \rightarrow$  channel bandwidth

$S \rightarrow$  Signal Power

$N \rightarrow$  Total noise power

$$\text{w.k.t Signal Power} = \int_{-B}^B \text{Power spectral density}$$

For white noise, PSD is  $\frac{N_0}{2}$ . Hence

$$N = \int_{-B}^B \frac{N_0}{2} df = \frac{N_0}{2} [f]_{-B}^B = \frac{N_0}{2} [B+B]$$

Bandwidth - S/N Trade off:

→ channel capacity of the Gaussian channel is given by,

$$C = B \log_2 \left[ 1 + \frac{S}{N} \right] \text{ bits/sec}$$

From the above equation, it is clear that the channel capacity depends on two factors.

(i) Band width

(ii) signal to noise ratio.

Noiseless channel has infinite capacity:

If there is no noise in the channel, then  $N = 0$ . Hence  $S/N = \infty$ . Such a channel is called noiseless channel. Then capacity of such a channel will be

$$C = B \log_2 [1 + \infty] = \infty.$$

Thus the noiseless channel has infinite capacity.

Infinite bandwidth channel has limited capacity:

→ Now if the bandwidth 'B' is infinite,

the channel capacity is limited. This is because, as bandwidth increases, noise power ( $N$ ) also increases.

Noise Power is given as,

$$N = N_0 B.$$

★ Due to this increase in noise power,  $S/N$  ratio decreases. Hence even if  $B$  approaches infinity, Capacity does not approach infinity.

As  $B \rightarrow \infty$ , Capacity approaches an upper limit. It is given as,

$$\lim_{B \rightarrow \infty} (C) = \frac{S}{N} \log e$$

$$\boxed{\lim_{B \rightarrow \infty} (C) = 1.44 \frac{S}{N}}$$

## Chapter 4

### Digital Transmission

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#### Line Coding

- Converting a string of 1's and 0's (digital data) into a sequence of signals that denote the 1's and 0's.
- For example a high voltage level (+V) could represent a "1" and a low voltage level (0 or -V) could represent a "0".

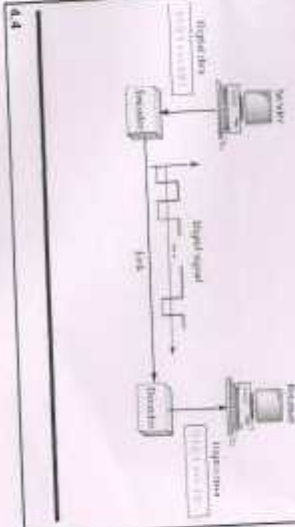
#### 4-1 DIGITAL-TO-DIGITAL CONVERSION

In this section, we see how we can represent digital data by using digital signals. The conversion involves three techniques: line coding, block coding, and scrambling. Line coding is always needed; block coding and scrambling may or may not be needed.

##### Topics discussed in this section:

- Line Coding
- Line Coding Schemes
- Block Coding
- Scrambling

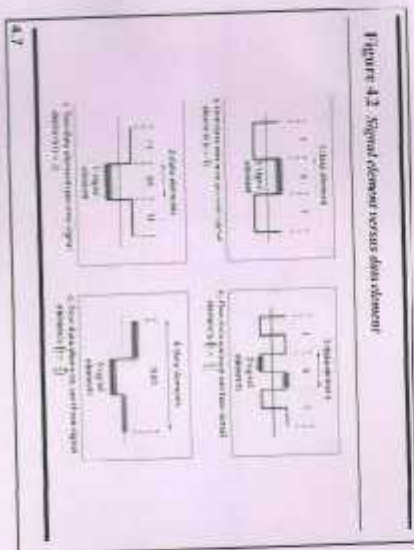
Figure 4.1 Line coding and decoding



### Mapping Data symbols onto Signal levels

- A data symbol (or element) can consist of a number of data bits:
  - 1, 0 or
  - 11, 10, 01, ....
- A data symbol can be coded into a single signal element or multiple signal elements
  - 1 -> +V, 0 -> -V
  - 1 -> +V and -V, 0 -> -V and +V
- The ratio 'r' is the number of data elements carried by a signal element.

Figure 4.2 Signal element versus data element



### Relationship between data rate and signal rate

- The data rate defines the number of bits sent per sec - bps. It is often referred to the bit rate.
- The signal rate is the number of signal elements sent in a second and is measured in bauds. It is also referred to as the modulation rate.
- Goal is to increase the data rate whilst reducing the baud rate.

### Data rate and Baud rate

- The baud or signal rate can be expressed as:

$$S = C \times N \times 1/r \text{ bauds}$$

where N is data rate

c is the case factor (worst, best & avg.)

r is the ratio between data element & signal element

**Example 4.1**

A signal is carrying data in which one data element is encoded as one signal element ( $r = 1$ ). If the bit rate is 100 kbps, what is the average value of the baud rate if  $c$  is between 0 and 1?

**Solution**

We assume that the average value of  $c$  is  $1/2$ . The baud rate is then

$$N_{baud} = N_{bit} \times \frac{1}{c} = \frac{1}{2} \times 100,000 = \frac{1}{2} \times 2\pi,000 = 50 \text{ kbaud}$$

4.9

**Note**

Although the actual bandwidth of a digital signal is infinite, the effective bandwidth is finite.

4.10

**Example 4.2**

The maximum data rate of a channel (see Chapter 3) is  $N_{max} = 2 \times B \times \log_2 L$  (defined by the Nyquist formula). Does this agree with the previous formula for  $N_{max}$ ?

**Solution**

A signal with  $L$  levels actually can carry  $\log_2 L$  bits per level. If each level corresponds to one signal element and we assume the average case ( $c = 1/2$ ), then we have

$$N_{max} = \frac{1}{2} \times B \times c = \frac{1}{2} \times B \times \log_2 L$$

4.11

Considerations for choosing a good signal element referred to as line encoding

- Baseline wandering - a receiver will evaluate the average power of the received signal (called line baseline) and use that to determine the value of the incoming data elements. If the incoming signal does not vary over a long period of time, the baseline will drift and thus cause errors in detection of incoming data elements.
- A good line encoding scheme will prevent long runs of fixed amplitude.

4.12



### Line encoding C/Cs

- Error detection - errors occur during transmission due to line impairments.
- Some codes are constructed such that when an error occurs it can be detected. For example: a particular signal transition is not part of the code. When it occurs, the receiver will know that a symbol error has occurred.

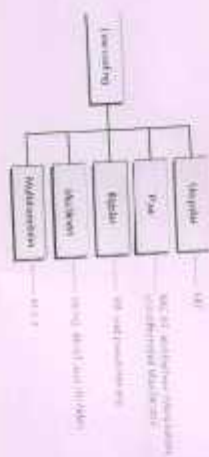
### Line encoding C/Cs

- Complexity - the more robust and resilient the code, the more complex it is to implement and the price is often paid in baud rate or required bandwidth.

### Line encoding C/Cs

- Noise and interference - there are line encoding techniques that make the transmitted signal "immune" to noise and interference.
- This means that the signal cannot be corrupted, it is stronger than error detection.

Figure 4.4 Line coding schemes



### Unipolar

- All signal levels are on one side of the time axis - either above or below
- NRZ - Non Return to Zero scheme is an example of this code. The signal level does not return to zero during a symbol transmission.
- Scheme is prone to baseline wandering and DC components. It has no synchronization or any error detection. It is simple but costly in power consumption.

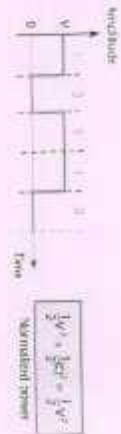
4.21

### Polar - NRZ

- The voltages are on both sides of the time axis.
- Polar NRZ scheme can be implemented with two voltages. E.g. +V for 1 and -V for 0.
- There are two versions:
  - NRZ - level (NRZ-L) - positive voltage for one symbol and negative for the other
  - NRZ - inversion (NRZ-I) - the change or lack of change in polarity determines the value of a symbol. E.g. a "1" symbol inverts the polarity a "0" does not.

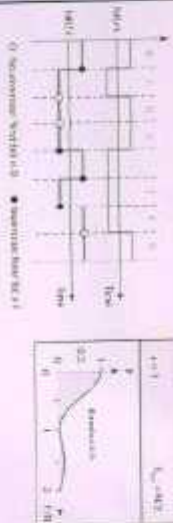
4.23

Figure 4.5 Unipolar NRZ scheme



4.25

Figure 4.6 Polar NRZ-L and NRZ-I schemes



4.26

1/1/2003

**Note**

In NRZ-L the level of the voltage determines the value of the bit.  
In NRZ-I the inversion or the lack of inversion determines the value of the bit.

4.25

**Note**

NRZ-L and NRZ-I both have an average signal rate of  $N/2$  Bd.

4.26

**Note**

NRZ-L and NRZ-I both have a DC component problem and baseline wandering, it is worse for NRZ-L. Both have no self synchronization & no error detection. Both are relatively simple to implement.

4.27

**Example 4.4**

A system is using NRZ-I to transfer 1-Mbps data. What are the average signal rate and minimum bandwidth?

**Solution**

The average signal rate is  $S = c \times N \times R = 1/2 \times N \times I = 500$  kbaud. The minimum bandwidth for this average baud rate is  $B_{\min} = S = 500$  kHz.

Note:  $c = 1/2$  for the avg. case as worst case is 1 and best case is 0.

4.28

### Polar - RZ

- The Return to Zero (RZ) scheme uses three voltage values, +, 0, -.
- Each symbol has a transition in the middle. Either from high to zero or from low to zero.
- This scheme has more signal transitions (two per symbol) and therefore requires a wider bandwidth.
- No DC components or baseline wandering.
- Self synchronization - transition indicates symbol value.
- More complex as it uses three voltage levels. It has no error detection capability.

4.29

Figure 4.7 Polar RZ scheme



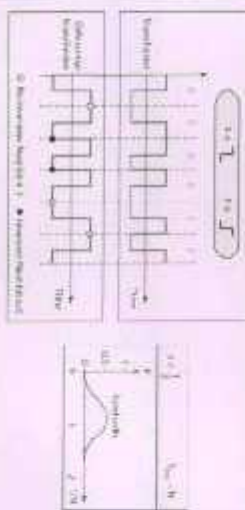
4.30

### Polar - Biphasic: Manchester and Differential Manchester

- Manchester coding consists of combining the NRZ-L and RZ schemes.
- Every symbol has a level transition in the middle, from high to low or low to high. Uses only two voltage levels.
- Differential Manchester coding consists of combining the NRZ-L and RZ schemes.
- Every symbol has a level transition in the middle, but the level at the beginning of the symbol is determined by the symbol value. One symbol causes a level change the other does not.

4.31

Figure 4.8 Polar biphasic: Manchester and differential Manchester schemes



4.32

6/1/2005

**Note**

In Manchester and differential Manchester encoding, the transition at the middle of the bit is used for synchronization.

4.33

**Note**

The minimum bandwidth of Manchester and differential Manchester is 2 times that of NRZ. There is no DC component and no baseline wandering. None of these codes has error detection.

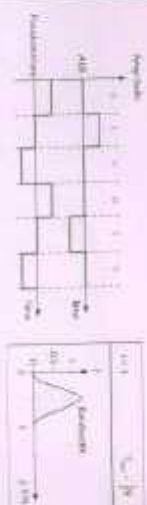
4.34

**Bipolar - AMI and Pseudoternary**

- Code uses 3 voltage levels:  $+V$ ,  $0$ ,  $-V$  to represent the symbols (note not transitions to zero as in RZ).
- Voltage level for one symbol is at  $+V$  and the other alternates between  $+V$  and  $-V$ .
- Bipolar Alternate Mark Inversion (AMI) - the  $+V$  symbol is represented by zero voltage and the  $-V$  symbol alternates between  $+V$  and  $-V$ .
- Pseudoternary is the reverse of AMI.

4.35

Figure 4.9 Bipolar schemes: AMI and pseudoternary



4.36

1/1/2002

### Bipolar C/Cs

- It is a better alternative to NRZ.
- Has no DC component or baseline wandering.
- Has no self synchronization because long runs of "0"s results in no signal transitions.
- No error detection.

### Code C/Cs

- Now we have  $2^m$  symbols and  $L^n$  signals.
- If  $2^m > L^n$  then we cannot represent the data elements, we don't have enough signals.
- If  $2^m = L^n$  then we have an exact mapping of one symbol on one signal.
- If  $2^m < L^n$  then we have more signals than symbols and we can choose the signals that are more distinct to represent the symbols and therefore have better noise immunity and error detection as some signals are not valid.

### Multilevel Schemes

- In these schemes we increase the number of data bits per symbol thereby increasing the bit rate.
- Since we are dealing with binary data we only have 2 types of data element a 1 or a 0.
- We can combine the 2 data elements into a pattern of " $m$ " elements to create " $2^m$ " symbols.
- If we have  $L$  signal levels, we can use " $n$ " signal elements to create  $L^n$  signal elements.

### Note

In *mBnL* schemes, a pattern of  $m$  data elements is encoded as a pattern of  $n$  signal elements in which  $2^m \leq L^n$ .

### Representing Multilevel Codes

- We use the notation  $mBnL$ , where  $m$  is the length of the binary pattern,  $n$  represents the length of the signal pattern and  $L$  the number of levels.
- 1 = B binary, L = T for 3 ternary, L = Q for 4 quaternary.

### Redundancy

- In the 2B1Q scheme we have no redundancy and we see that a DC component is present.
- If we use a code with redundancy we can decide to use only "0" or "-" weighted codes (more +s than -s in the signal element) and invert any code that would create a DC component. E.g.  $+00++ \rightarrow -00++$
- Receiver will know when it receives a "-" weighted code that it should invert it as it doesn't represent any valid symbol.

Figure 4.10 Multilevel 2B1Q scheme

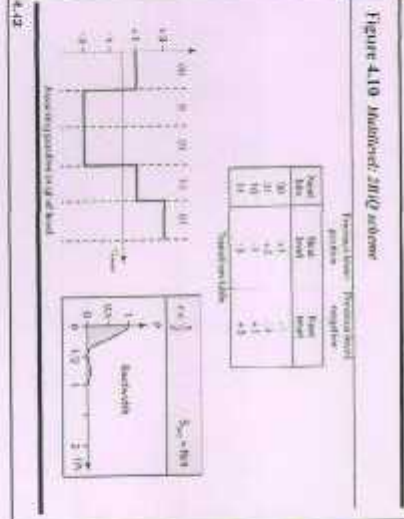
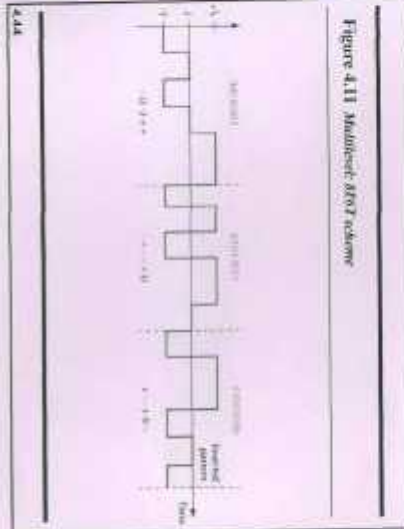


Figure 4.11 Multilevel 8B6T scheme



V/12/2402

### Multilevel using multiple channels

- In some cases, we split the signal transmission up and distribute it over several links.
- The separate segments are transmitted simultaneously. This reduces the signalling rate per link  $\rightarrow$  lower bandwidth.
- This requires all bits for a code to be stored.
- xD: means that we use 'x' links.
- yyyz: We use 'y' levels of modulation where 'yyy' represents the type of modulation (e.g. pulse amplit. mod. PAM).
- Codes are represented as: xD-yyyz

### Multitransition Coding

- Because of synchronization requirements, we force transitions. This can result in very high bandwidth requirements  $\rightarrow$  more transitions than are bits (e.g. mid bit transition with inversion).
- Codes can be created that are differential at the bit level forcing transitions at bit boundaries. This results in a bandwidth requirement that is equivalent to the bit rate.
- In some instances, the bandwidth requirement may even be lower, due to repetitive patterns resulting in a periodic signal.

Figure 4.12 Multilevel 4D-PAM scheme

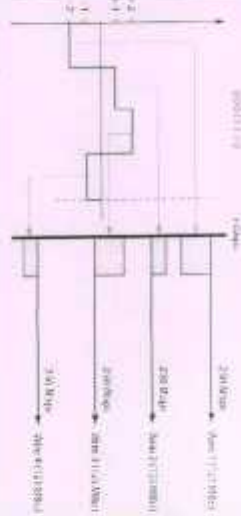
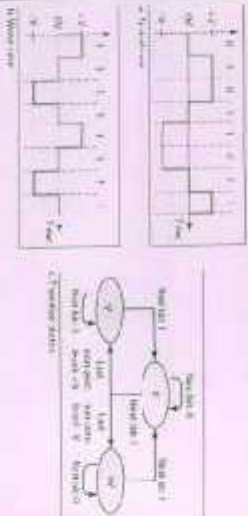


Figure 4.13 Multitransition MZT-3 scheme



## MLT-3

- Signal rate is same as NRZ-1
- But because of the resulting bit pattern, we have a periodic signal for worst case bit pattern: 1111
- This can be approximated as an analog signal a frequency  $1/4$  the bit rate!

## Block Coding

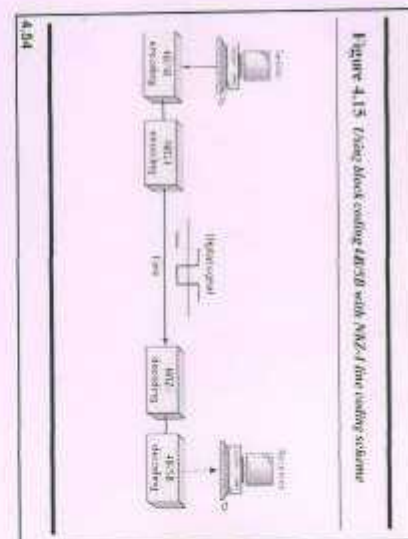
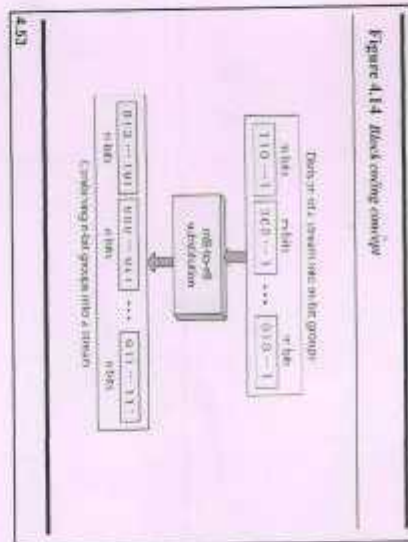
- For a code to be capable of error detection, we need to add redundancy, i.e., extra bits to the data bits.
- Synchronization also requires redundancy - transitions are important in the signal flow and must occur frequently.
- Block coding is done in three steps: division, substitution and combination.
- It is distinguished from multilevel coding by use of the slash -  $kB/nB$ .
- The resulting bit stream prevents certain bit combinations that when used with line encoding would result in DC components or poor sync. quality.

Table 4.1 Summary of line coding schemes

Encoding Technique	Signal Representation	Redundancy (overhead)	Characteristics
Unipolar			
NRZ	$R = 2C$	$R = 2C$	Can't do self synchronization; long runs of 1s or 0s
NRZ-L	$R = 2C$	$R = 2C$	No self synchronization; long runs of 1s or 0s
NRZ-I	$R = 2C$	$R = 2C$	No self synchronization; long runs of 1s or 0s
Biphase			
AVB	$R = 1.5C$	$R = 1.5C$	Self synchronization; no DC; high bandwidth
2B1Q	$R = 1.5C$	$R = 1.5C$	Self synchronization; no DC; high bandwidth
Manchester			
4B/1Q	$R = 2.5C$	$R = 2.5C$	Self synchronization; no DC; high bandwidth
4B/2Q	$R = 2.5C$	$R = 2.5C$	Self synchronization; no DC; high bandwidth
MLT-3	$R = 1.5C$	$R = 1.5C$	No self synchronization; long runs of 1s or 0s

Note

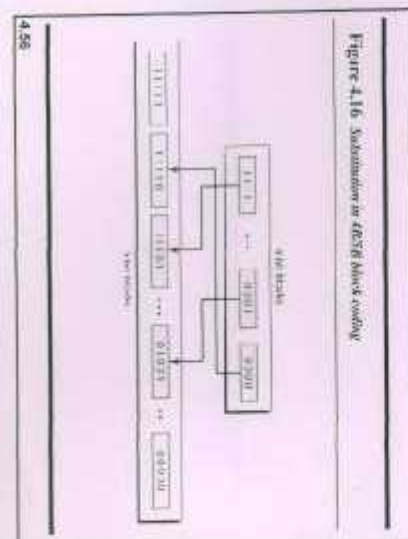
Block coding is normally referred to as  $mB/nB$  coding: it replaces each  $m$ -bit group with an  $n$ -bit group.



**Table 4.2 4B/5B mapping codes**

Hex Sequence	Binary Sequence	4B/5B Mapping	Hex Sequence
0000	00000	11111	0000
0001	00001	11110	0001
0010	00010	11101	0010
0011	00011	11100	0011
0100	00100	11011	0100
0101	00101	11010	0101
0110	00110	11001	0110
0111	00111	11000	0111
1000	01000	10111	1000
1001	01001	10110	1001
1010	01010	10101	1010
1011	01011	10100	1011
1100	01100	10011	1100
1101	01101	10010	1101
1110	01110	10001	1110
1111	01111	10000	1111

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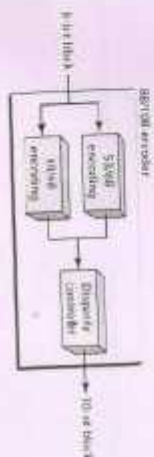


### Redundancy

- A 4 bit data word can have 24 combinations.
- A 5 bit word can have  $2^5 = 32$  combinations.
- We therefore have  $32 - 2^4 = 16$  extra words.
- Some of the extra words are used for control/signalling purposes.

4.57

Figure 4.17 8B/10B block encoding



4.59

### Example 4.5

We need to send data at a 1-Mbps rate. What is the minimum required bandwidth, using a combination of 4B/5B and NRZ-I or Manchester coding?

*Solution*

First 4B/5B block coding increases the bit rate to 1.25 Mbps. The minimum bandwidth using NRZ-I is  $N/2$  or 625 MHz. The Manchester scheme needs a minimum bandwidth of 1.25 MHz. The first choice needs a lower bandwidth, but has a DC component problem; the second choice needs a higher bandwidth, but does not have a DC component problem.

4.58

### More bits - better error detection

- The 8B/10B block code adds more redundant bits and can thereby choose code words that would prevent a long run of a voltage level that would cause DC components.

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## Scrambling

- The best code is one that does not increase the bandwidth for synchronization and has no DC components.
- Scrambling is a technique used to create a sequence of bits that has the required  $c/f_c$  for transmission - self clocking, no low frequencies, no wide bandwidth.
- It is implemented at the same time as encoding, the bit stream is created on the fly.
- It replaces 'unfriendly' runs of bits with a violation code that is easy to recognize and removes the unfriendly  $c/f_c$ .

4.51

For example: B8ZS substitutes eight consecutive zeros with 000VB0VB. The V stands for violation, it violates the line encoding rule

B stands for bipolar, it implements the bipolar line encoding rule

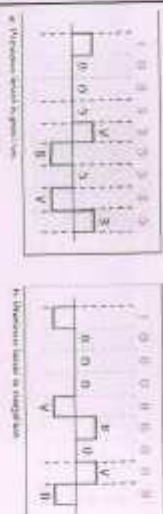
4.63

Figure 4.18 ANT spot with ascending



4.53

Figure 4.19: Two cases of B&Z scrambling in Belgium



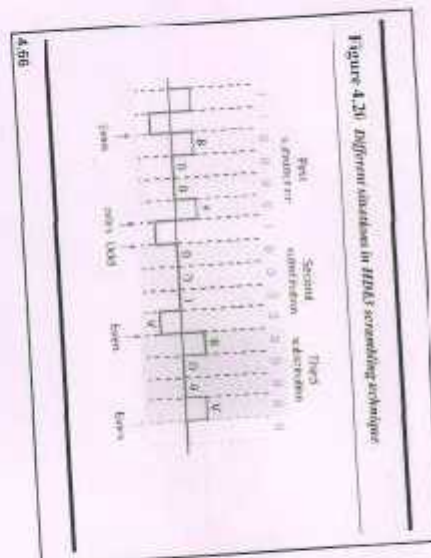
400

4.35

HDB3 substitutes four consecutive zeros with 000V or B00V depending on the number of non-zero pulses after the last substitution.

If # of non zero pulses is even the substitution is B00V to make total # of non zero pulses even.

If # of non zero pulses is odd the substitution is 000V to make total # of non zero pulses even.



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## Error Control Codes

→ Mechanism involves both error detection & error correction.

→ to improve data quality.

Two error control codes

\* Block codes

\* Convolutional codes.

### Terms

(i) Block (or) code word: → It consists of 'n' number of bits. This CW contains message bits & redundant bits.

(ii) Block length: The number of bits 'n' after coding.

(iii) Code rate:  $r = \frac{k}{n}$  (message bits) / (encoder o/p bits)  $0 < r < 1$

(iv) channel data rate: → It is the bit rate at the o/p of the encoder. If the bit rate at the i/p of the encoder is  $R_s$ , then

$$R_o = \left[ \frac{n}{k} \right] R_s$$

(v) Code Vectors:

0	0	0
0	0	1
0	1	0
0	1	1
1	0	0

(vi) Hamming distance

Eg:  $A = (111)$ ,  $B = (110)$ . The two code vectors differ in third bit. Therefore hamming distance b/w  $A$  &  $B$  is 'one'.

$$d = 1.$$

In 3 bit code vectors, Max. hamming distance is 3.

Eg:  $A = (111)$ ,  $B = (010)$ .

$$d = 2.$$

(vii) Minimum distance: ( $d_{min}$ ):

$$d_{min} = n - k + 1$$

(viii) weight of the code

The number of non-zero elements in the transmitted code vector is a code weight.

for eg:  $A = 11010001$ ,

$$W(A) = 4$$

parity bits with condition  $n > k$ . Such a code formed is called  $(n, k)$  block codes.

Linear block Codes:

set in In channel encoder, a block of  $k$  message bits are encoded into a block of  $n$  bits by adding  $(n-k)$  number of

A block code ~~(n, k)~~ is said to be  $(n, k)$  linear block code, if it satisfies the

- (i) Linear Property → Let  $C_1$  &  $C_2$  be any two code words of  $n$  bits belonging to a set of  $(n, k)$  block code. If  $C_1 \oplus C_2$  is also an  $n$ -bit code word belonging to the same set of  $(n, k)$  block code, then such a code is called  $(n, k)$  linear block code.
- (ii) systematic Property

(i) Linear Property

Let us consider two code words of 'n' bits

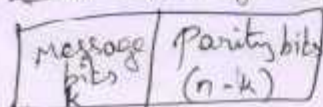
$C_1$  &  $C_2 \Rightarrow C_1 = 1111, C_2 = 0000$   
→ belonging to a set of block codes

$C_1 \oplus C_2 \rightarrow$  belonging to the same set of "

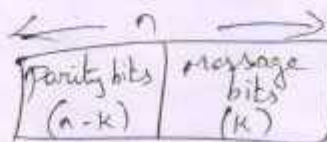
(ii) systematic Property

A linear block code is said to be systematic if  $k$  message bits appear either at the beginning or at the end of the code word.

← code word length =  $n$  bits

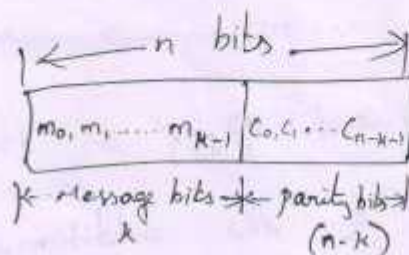


(or)



Matrix description of linear block codes:

Structure of code word



\* In a systematic linear block code, the first  $k$  bits of the code word are the Message bits i.e.

$$C_i = m_i \longrightarrow \textcircled{1}$$

where  $i = 1, 2, \dots, k$

\* The last  $(n-k)$  bits in the code word are Parity bits generated from the  $k$  Message bits according to some determined rule.

$$C_{k+1} = P_{11}m_1 + P_{21}m_2 + \dots + P_{k1}m_k$$

$n-k=1$   
 $n-k+1$   
 $n-k+2$

$$C_{k+2} = P_{12}m_1 + P_{22}m_2 + \dots + P_{k2}m_k$$

$\vdots$

$$C_n = P_{1,n-k}m_1 + P_{2,n-k}m_2 + \dots + P_{k,n-k}m_k$$

where  $C_{k+1}, C_{k+2}, \dots, C_n \rightarrow$  Corresponding CW bits.  
 $m_1, m_2, \dots, m_k \rightarrow$  " msg bits.

The coefficient  $P_{ij}$ 's in the above equation are 0's & 1's & the additional operations are

$$1 \oplus 1 = 0$$

$$1 \oplus 0 = 1$$

$$0 \oplus 1 = 1$$

$$0 \oplus 0 = 0$$

\* Combining equation ① & ② and write them in matrix form.

$$[C_1, C_2, \dots, C_n] = [m_1, m_2, \dots, m_k] \begin{bmatrix} 1 & 0 & \dots & 0 & p_{11} & p_{12} & \dots & p_{1n} \\ 0 & 1 & \dots & 0 & p_{21} & p_{22} & \dots & p_{2n} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 & p_{k1} & p_{k2} & \dots & p_{kn} \end{bmatrix}$$

It can be written as,

$$C = M G$$

$G \rightarrow$  Generator Matrix of order  $(k \times n)$

\* The Generator Matrix is of the form

$$G = [I_k : P_k]_{k \times n}$$

$I_k \rightarrow$  Identity Matrix

If  $k=3$ , then  $I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$

## Parity check matrix (H)

→ For every block code, there is a  $H (q \times n)$ .

It is defined as,

$$H = \left[ P^T : I_q \right]_{q \times n} \rightarrow \textcircled{1}$$

$P^T \rightarrow$  transpose of  $P$  sub-matrix.

The  $P$  sub matrix is defined as,

$$P = \begin{bmatrix} P_{11} & P_{12} & P_{13} & \dots & P_{1q} \\ P_{21} & P_{22} & P_{23} & \dots & P_{2q} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ P_{k1} & P_{k2} & P_{k3} & \dots & P_{kq} \end{bmatrix}_{k \times q}$$

The transpose of this sub matrix is

$$P^T = \begin{bmatrix} P_{11} & P_{12} & \dots & P_{1k} \\ P_{21} & P_{22} & \dots & P_{2k} \\ \vdots & \vdots & \ddots & \vdots \\ P_{q1} & P_{q2} & \dots & P_{qk} \end{bmatrix}_{q \times k}$$

$$H_{q \times n} = \begin{bmatrix} P_{11} & P_{12} & \dots & P_{1k} & : & 1 & 0 & \dots & 0 \\ P_{21} & P_{22} & \dots & P_{2k} & : & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots & : & \vdots & \vdots & \ddots & \vdots \end{bmatrix}$$

### 1) Syndrome Decoding:

→ It is one of the methods used to correcting errors in linear block coding.

→ Let the transmitted code vector be 'X' and corresponding received code vector be 'Y'. Then, it can be expressed as,

$X = Y$  if there are no transmission errors.

$X \neq Y$  if there are errors created during transmission.

W.K.T with every  $(n, k)$  linear block codes, there exists a Parity check matrix (H) of size  $q \times n$  is given as.

$$H = [P^T : I_q]_{q \times n}$$

→ The transpose of the above matrix can be obtained by interchanging the rows & the columns, i.e

$$H^T = \begin{bmatrix} P \\ \vdots \\ I_q \end{bmatrix}_{n \times q}$$

Important Property used in syndrome decoding

→ The transpose of parity check matrix ( $H^T$ ) has very important Properties as follows.

### Definition of Syndrome (s):

When some errors are present in received vector  $y$ , then it will not be from valid code vectors and it will not satisfy the property of syndrome ~~error~~ as in <sup>above</sup> equations.

$XH^T = (000 \dots 0)$ . This shows that whenever  $yH^T$  is non-zero, some errors are present in  $y$ . The non zero output of the product  $yH^T$  is called syndrome and it is used to detect the errors in  $y$ .

Syndrome represented by 's' and can be written as,

$$S = yH^T$$

### Detecting error with the help of Syndrome and error vector (E):

→ The non-zero elements of 's' represent error in the output. When all elements of 's' are zero, the two cases are possible.

- (i) No error in the op and  $y = x$ .
- (ii)  $y$  is some other valid code vector other than  $x$ . This means the transmission errors are undetectable.

Syndrome vector (S) Vs Error Vector (E)

w.k.t, the syndrome Vector is given as,

$$S = YH^T$$

By replacing  $Y = (X \oplus E)$

$$S = (X \oplus E)H^T$$

$$= (XH^T \oplus EH^T) \quad (\text{since, } XH^T = 0)$$

$$= 0 \oplus EH^T$$

$$\therefore S = EH^T$$

The above equation clearly shows that, the syndrome depends only on the error pattern and not upon a particular message.

The other linear block codes are,

- (i) single parity check bit code,
- (ii) Repeated codes
- (iii) Hadamard code, &
- (iv) Dual code.

Ex The Parity matrix  $P$  for a systematic  $(6,3)$  linear block code is given as,  $[P] = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix}$ . Find all the possible code vectors.

Ans

(i)  $n=6, k=3$  for  $(6,3)$  linear block code.

Since  $k=3$ , the possible message vectors are  $2^k = 2^3 = 8$ , which is given as,

S.No	Bits of Message Vector		
	$m_1$	$m_2$	$m_3$
1	0	0	0
2	0	0	1
3	0	1	0
4	0	1	1
5	1	0	0
6	1	0	1
7	1	1	0
8	1	1	1

(ii) The ~~Parity~~ parity matrix as,

$$[P] = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix}$$

To find all the possible code vectors.

The equation for code vectors are:

$$C = M G$$

W.K.T

$$G = [I_k : P]$$

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

$$= \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix}$$

$$\therefore [C_1 \ C_2 \ C_3] = [m_1 \ m_2 \ m_3] \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix}$$

which means,

$$C_1 = m_1 \oplus m_3$$

$$C_2 = m_1 \oplus m_2 \oplus m_3$$

$$C_3 = 0$$

Message Vector			Code Vector		
$m_1$	$m_2$	$m_3$	$C_1$	$C_2$	$C_3$
0	0	0	0	0	0
0	0	1	1	1	0
0	1	0	0	1	0
0	1	1	1	0	0
1	0	0	1	1	0
1	0	1	0	0	0
1	1	0	1	0	0
1	1	1	0	1	0

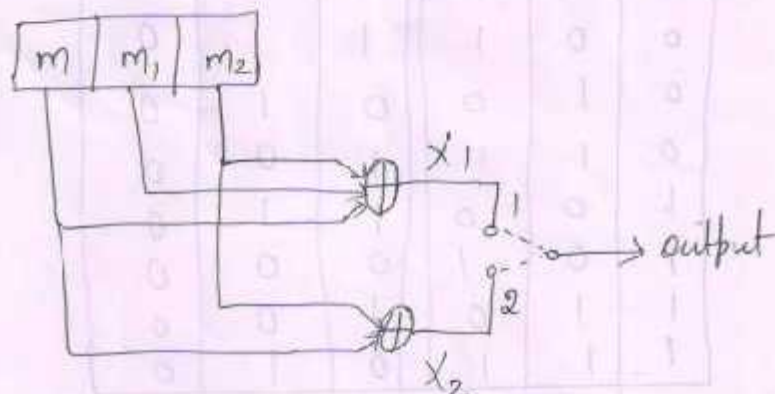
- ② The parity matrix  $P$  for a systematic  $(6,3)$  linear block code is given as,  $[P] = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}$ . Find all the possible code vectors.

### Convolutional Codes:

A Convolutional Coding is done by combining the fixed number of input bits. The input bits are stored in the fixed length shift register and they are combined with the help of modulo-2 adders. This operation is equivalent to the binary convolution and hence it is called convolutional coding.

Encoder:

Consider the following encoder diagram



$$X_1 = m \oplus m_1 \oplus m_2$$

$$X_2 = m \oplus m_2$$

→ The output switch first samples  $X_1$  and then  $X_2$ .  
The shift register then shifts contents of  $m_1$  to  $m_2$  and contents of  $m$  to  $m_1$ .

→ next i/p bit is then taken & stored in  $m$ .

→ Again  $X_1$  and  $X_2$  are generated according to this new combination of  $m$ ,  $m_1$  &  $m_2$ .

→ The o/p switch then samples  $X_1$  & then  $X_2$ .

Thus the o/p bit stream for successive input bits will be,

For every input message bit, 2 output bits  $x_1$  and  $x_2$  are transmitted.

$\therefore$  message bit  $k=1$  which results in the 2 output bits i.e.  $n=2$ .

Code Rate: The code rate of this encoder is

$$r = \left[ \frac{k}{n} \right] = \left[ \frac{1}{2} \right]$$

From the above figure.

First shift  $\rightarrow$  message bit is entered in position 'm'

Second shift  $\rightarrow$  " " " " 'm'

Third shift  $\rightarrow$  " " " " 'm'

Fourth shift  $\rightarrow$  the message bit is discarded.

Constraint Length: (K)

It is defined as, "the number of shifts over which a single message bit can influence the encoder output".

From the above figure encoder diagram

encoder, a single message bit influences encoder for three successive shifts.

At the fourth shift, the message bit is lost and it has no effect on the output.

The constraint length ( $K$ ) can also be expressed as,

$$K = (n \times m)$$

where  $n \rightarrow$  number of o/p bits for every i/p bit  
 $m \rightarrow$  " " storage elements in the shift register

$$\therefore n = 2, m = 3$$

$$\therefore K = 2 \times 3 = 6 \text{ bits.}$$

Dimension of the Code:

The dimension of the code is given by  $n$  and  $k$ .

where  $k =$  number of message bits taken at a time.

$n =$  encoded output bits for 1 message bit.

Hence, dimension of code is  $(n, k) = (2, 1)$ .

Convolutional encoder has two approaches.

Convolutional Encoder has two approaches

(i) Time Domain Approach to analysis of Convolutional Encoder:

\* Let us consider the impulse response of the adder with sequences  $\{g_0^{(1)}, g_1^{(1)}, g_2^{(1)} \dots g_m^{(1)}\}$  which generates  $X_1$  as shown in above figure.

\* Similarly, the impulse response of the adder with sequences  $\{g_0^{(2)}, g_1^{(2)}, g_2^{(2)} \dots g_m^{(2)}\}$  which generates  $X_2$  as shown in above figure.

→ These impulse responses are also called as generator sequences of the code.

Let the incoming Message sequence be  $\{m_0, m_1, m_2 \dots\}$ .

→ The encoder generates the 2 output sequences  $X_1$  and  $X_2$ . These are obtained by convolving the message sequence with generator sequence.

The sequences  $X_1$  and  $X_2$  is given as,

$$X_1 = X_i^{(1)} = \left[ \sum_{l=0}^m g_l^{(1)} m_{i-l} \right] \text{ for}$$

The sequence  $x_2$  is given as . .

$$x_2 = x_i^{(2)} = \left[ \sum_{l=0}^m g_l^{(2)} m_{i-l} \right] \quad i=0,1,2,\dots,1>i, \quad i-1=0 \rightarrow \textcircled{2}$$

$\therefore$  output sequence is given by

$$\{x_i\} = \{x_0^{(1)} x_0^{(2)} x_1^{(1)} x_1^{(2)} x_2^{(1)} x_2^{(2)} x_3^{(1)} x_3^{(2)} \dots\}$$

Here

$$x_i^{(1)} = \{x_0^{(1)}, x_1^{(1)}, x_2^{(1)} \dots\}$$

$$x_i^{(2)} = \{x_0^{(2)}, x_1^{(2)}, x_2^{(2)} \dots\}$$

## (ii) Transform Domain Approach to analysis of Convolutional Encoder

The convolution of generating and message sequences takes place in the time domain approach.

It can be simplified by adding transformations to the sequences.

Let the impulse response be represented by,

$$g^{(1)}(p) = g_0^{(1)} p^0 + g_1^{(1)} p^1 + g_2^{(1)} p^2 + \dots + g_m^{(1)} p^m$$

$$g^{(2)}(p) = g_0^{(2)} p^0 + g_1^{(2)} p^1 + g_2^{(2)} p^2 + \dots + g_m^{(2)} p^m$$

→ with respect to the above generator polynomial the message polynomial is given as,

$$m(p) = m_0 + m_1 p + m_2 p^2 + \dots + m_{L-1} p^{L-1}$$

where,  $L$  = length of message sequence

The output of adders is given as,

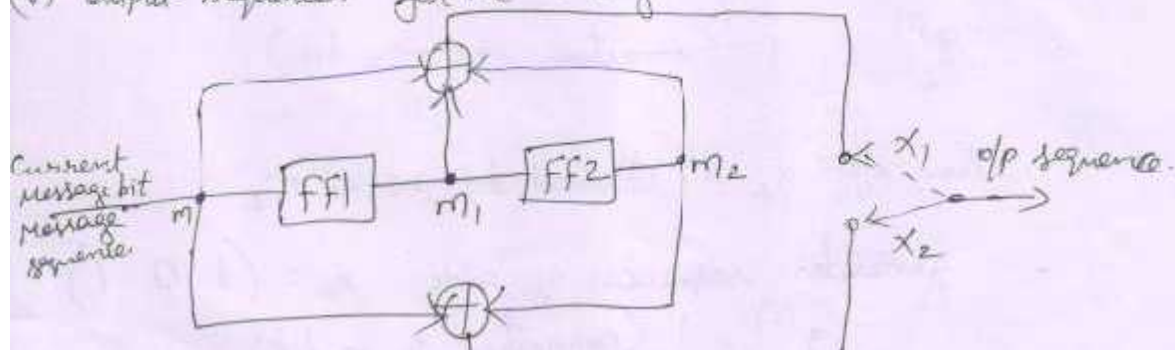
$$x^{(1)}(p) = g^{(1)}(p) m(p)$$

$$x^{(2)}(p) = g^{(2)}(p) m(p)$$

### Problem ①

for the convolutional encoder, determine the following

- (i) dimension of the code
- (ii) code rate
- (iii) constraint length
- (iv) Generating sequences (impulse responses)
- (v) output sequences for the message sequence  $m = 10011$



Given:

$n=2$  (no. of outputs)

$k=1$  (no. of i/p bit)

soln:

(i) Dimension of the code

$$(n, k) = (2, 1)$$

(ii) code rate

$$r = \frac{k}{n} = \frac{1}{2}$$

(iii) Constraint length

$$K = 3 \text{ bits (or)} n \times M = 2 \times 3 = 6 \text{ bits.}$$

(iv) Generator sequences:

The output  $x_1$  is obtained by all  $m, m_1, m_2$ .

$\therefore$  generator sequences of adder  $x_1 = (1 \ 1 \ 1)$

$$g_0^{(1)} = 1 \text{ (connection of } m \text{ bit)}$$

$$g_1^{(1)} = 1 \text{ (connection of } m_1 \text{ bit)}$$

$$g_2^{(1)} = 1 \text{ (connection of } m_2 \text{ bit)}$$

The o/p  $x_2$  is obtained by  $m$  &  $m_2$

$\therefore$  generator sequences of adder  $x_2 = (1 \ 0 \ 1)$

$$g_0^{(2)} = 1 \text{ (connection of } m \text{ bit)}$$

$$g_2^{(2)} = 1 \text{ (connection of } m_2 \text{ bit).}$$

(v) Output sequences for Message sequence  $m = \overset{m_0 m_1 m_2 m_3 m_4}{10011}$

o/p due to adder 1

$$X_1 = X_i^{(1)} = \left[ \sum_{l=0}^m g_l^{(1)} m_{i-l} \right] \text{ for } i=0, 1, 2, \dots, i > i, \quad i-l=0$$

$$\text{when } i=0, \quad X_0^{(1)} = \sum_{l=0}^m g_l^{(1)} m_{0-l}$$

$$= g_0^{(1)} m_0 = 1 \times 1 = 1.$$

$$i=1, \quad X_1^{(1)} = \sum_{l=0}^m g_l^{(1)} m_{1-l} = g_0^{(1)} m_1 \oplus g_1^{(1)} m_0$$

$$= (1)(0) \oplus (1)(1)$$

$$= 0 \oplus 1 = 1$$

$$i=2, \quad X_2^{(1)} = \sum_{l=0}^m g_l^{(1)} m_{2-l} = g_0^{(1)} m_2 \oplus g_1^{(1)} m_1 \oplus g_2^{(1)} m_0$$

$$= (1)(1) \oplus (1)(0) \oplus (1)(1)$$

$$= 1 \oplus 0 \oplus 1$$

$$= 0$$

$$i=3, \quad X_3^{(1)} = \sum_{l=0}^m g_l^{(1)} m_{3-l} = g_0^{(1)} m_3 \oplus g_1^{(1)} m_2 \oplus g_2^{(1)} m_1 \oplus g_3^{(1)} m_0$$

$$i=4, x_4^{(1)} = 1$$

$$i=5, x_5^{(1)} = 0$$

$$i=6, x_6^{(1)} = 1$$

111|001

The o/p of adder 1 is 1100101

o/p due to adder 2

$$x_2 = x_i^{(2)} = \sum_{l=0}^m g_l^{(2)} m_{i-l} \quad \text{for } i=0, 1, 2, \dots, 1 \geq i, \quad i-l=0$$

$$\text{when } i=0, x_0^{(2)} = \sum_{l=0}^m g_l^{(2)} m_{0-l}$$

$$= g_0^{(2)} m_0 = 1$$

$$i=1, x_1^{(2)} = 0$$

$$i=2, x_2^{(2)} = 0$$

$$i=3, x_3^{(2)} = 1$$

$$i=4, x_4^{(2)} = 0$$

$$i=5, x_5^{(2)} = 1$$

$$i=6, x_6^{(2)} = 1$$

1011111

The o/p of adder 2 is 1001011

$\therefore$  The overall output is obtained by combining output of adder 1 with adder 2 i.e. 2 first bits, 2 second bits, & so on. (11, 10, 11, 11, 01, 01, 11)

② Solve the above eg. Problem of Convolutional encoder using Transform domain Approach.

Generator sequences:

The o/p  $X_1$  is obtained by  $m, m_1, m_2$ .  
 $\therefore$  generator sequences of adder  $X_1 = (1 \ 1 \ 1) = g_i^{(1)}$

The polynomial can be obtained as,

$$g^{(1)}(p) = g_0^{(1)} p^0 + g_1^{(1)} p^1 + g_2^{(1)} p^2 \\ = 1 + p + p^2$$

The output  $X_2$  is obtained by  $m$  &  $m_2$ .

$\therefore$  generator sequences of adder  $X_2 = (1 \ 0 \ 1) = g_i^{(2)}$

The polynomial can be obtained as,

$$g^{(2)}(p) = g_0^{(2)} p^0 + g_1^{(2)} p^1 + g_2^{(2)} p^2 \\ = 1 + p^2$$

Output sequences for message sequence  $m = 10011$

with message polynomial  $= 1 + p^3 + p^4$

Output due to adder 1:

(1) (1) (1) (1) (1)

$$= (1 + p + p^2) (1 + p^3 + p^4)$$

$$= 1 + p^3 + p^4 + p + p^4 + p^5 + p^2 + p^5 + p^6$$

$$= 1 + p + p^2 + p^3 + p^4 + p^5 + p^6$$

$$= 1 + p + p^2 + p^3 + p^6 + p^4 + p^4$$

$$= (1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 1)$$

output due to adder 2:

$$X^{(2)}(p) = g^{(2)}(p) m(p)$$

$$= (1 + p^2) (1 + p^3 + p^4)$$

$$= 1 + p^3 + p^4 + p^2 + p^5 + p^6$$

$$= 1 + p^2 + p^3 + p^4 + p^5 + p^6$$

$$= (1 \ 0 \ 1 \ 1 \ 1 \ 1 \ 1)$$

The overall output is obtained by combining output of adder 1 with adder 2.

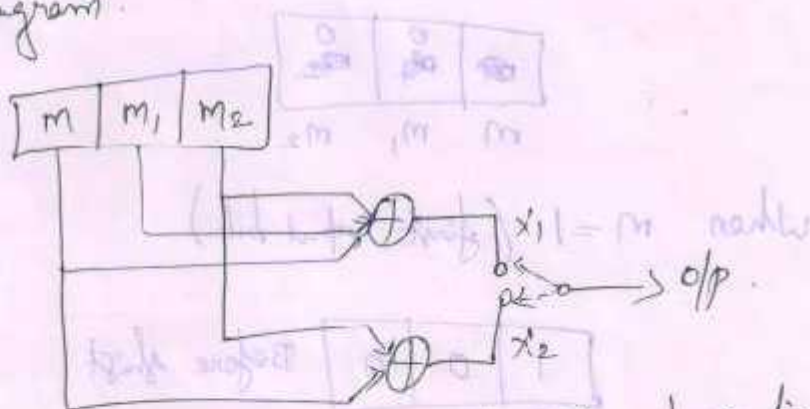
$$(1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 1) (1 \ 0 \ 1 \ 1 \ 1 \ 1 \ 1) = (11, 10, 11, 11, 01, 01, 11)$$

if repeated neglect it  
 $p^4$  &  $p^4$   
 two times  
 so neglect

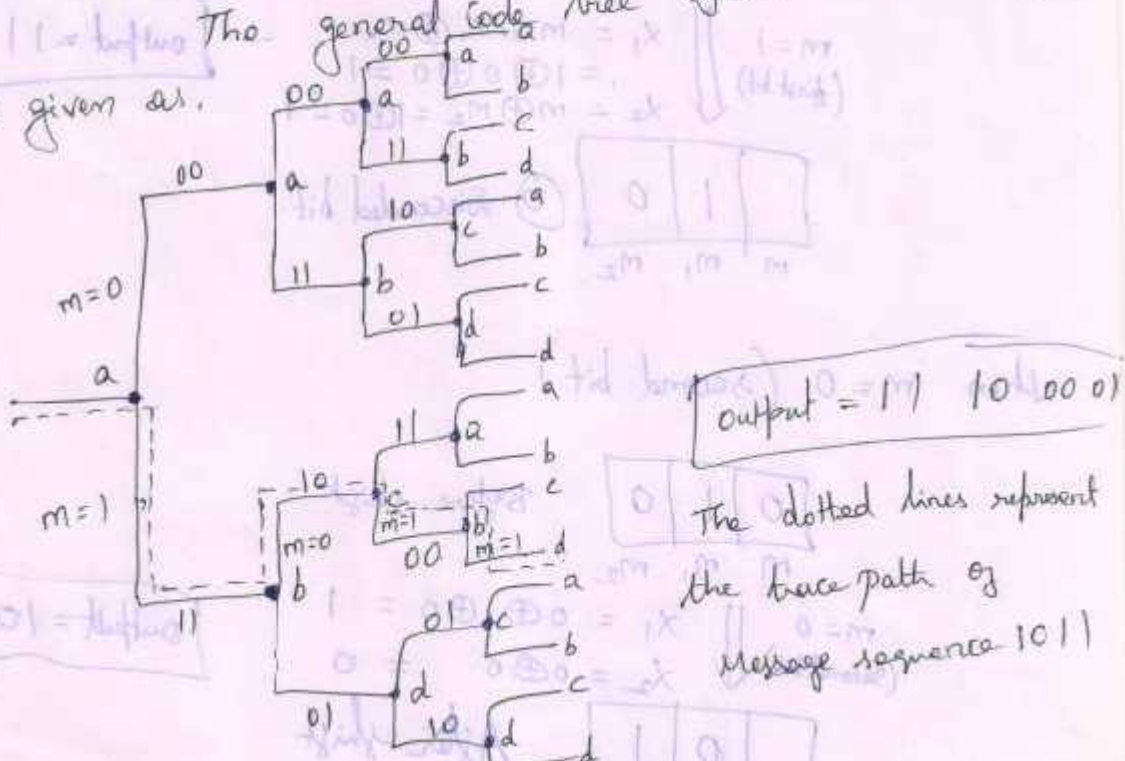
### Decoding Convolutional codes:

To decode a Convolutional Code, we consider the code tree.

Here we explain it with the convolutional encoder diagram.

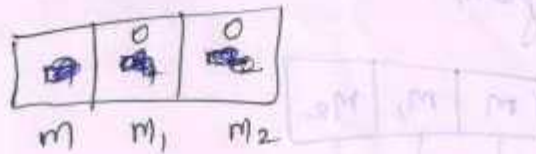


The general code tree for the above figure is given as.

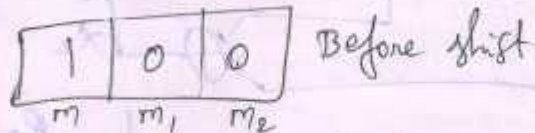


Let us consider the development of code tree for the input sequence  $m = 1011$

Initially assume  $m, m_1, m_2 = 00$

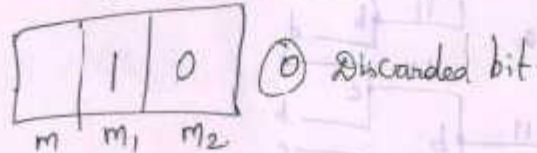


When  $m=1$  (first input bit)

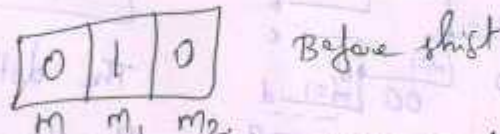


$$\begin{aligned}
 m=1 \text{ (first bit)} \quad & \begin{aligned} x_1 &= m \oplus m_1 \oplus m_2 \\ &= 1 \oplus 0 \oplus 0 = 1 \\ x_2 &= m \oplus m_2 = 1 \oplus 0 = 1 \end{aligned}
 \end{aligned}$$

output = 11



When  $m=0$  (second bit)



$$\begin{aligned}
 m=0 \text{ (second bit)} \quad & \begin{aligned} x_1 &= 0 \oplus 1 \oplus 0 = 1 \\ x_2 &= 0 \oplus 0 = 0 \end{aligned}
 \end{aligned}$$

output = 10





## State Diagram and its States:

→ From the above convolutional encoder diagram, the message bits  $m_1$  &  $m_2$  represents the state.

→ The input Message bit affects the state of the encoder as well as the outputs  $x_1$  &  $x_2$  during that state.

→ whenever new message bit is shifted to 'm', the contents of  $m_1$  and  $m_2$  define new state. And outputs  $x_1$  and  $x_2$  are also changed according to the new state  $m_1, m_2$  &  $m$ .

$m_2$	$m_1$	State of encoder
0	0	a
0	1	b
1	0	c
1	1	d

## State diagram:

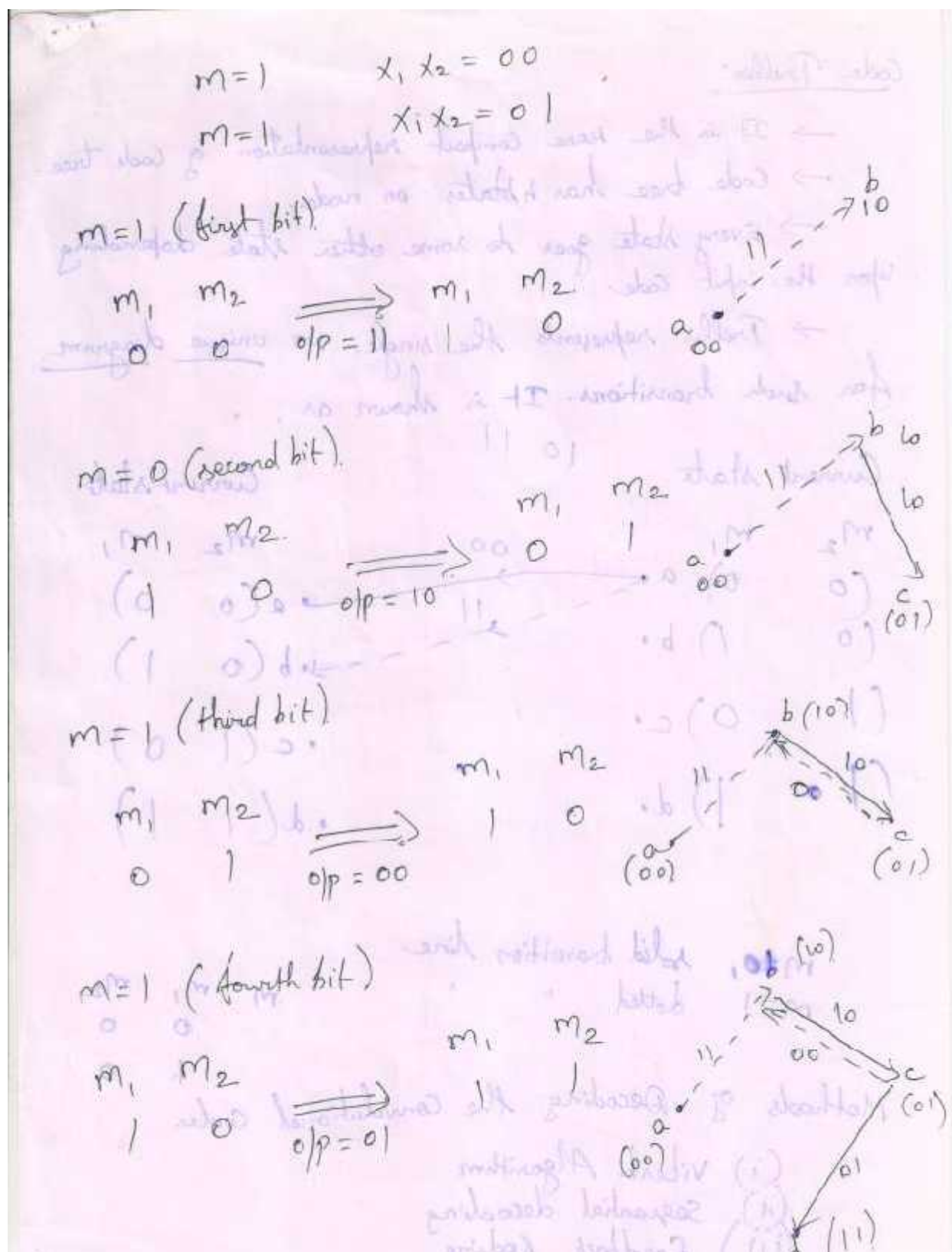
By combining the Current and next states, we can obtain the state diagram.

It can be easily drawn from the code tree.

Eg:

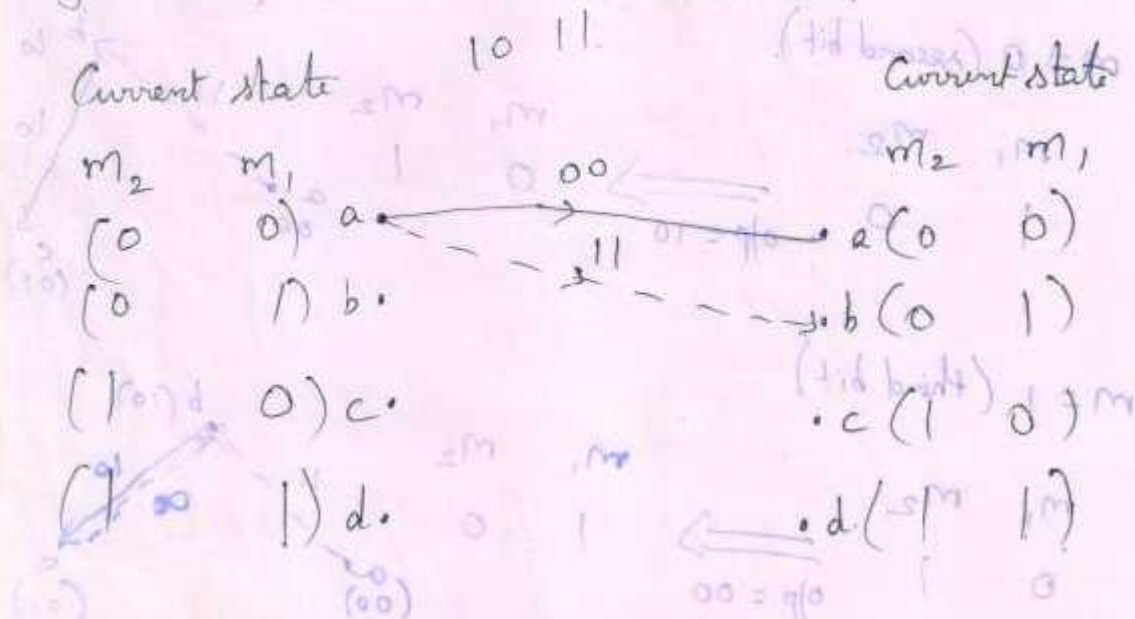
$m = 1011$

$x_1, x_2 = 11$



Code Trellis:

- It is the more compact representation of code tree.
- Code tree has 4 states on nodes.
- Every state goes to some other state depending upon the input code.
- Trellis represents the single, or unique diagram for such transitions. It is shown as,



$m=0$ , solid transition line  
 $m=1$ , dotted " "

Methods of Decoding the Convolutional Code.

- Viterbi Algorithm
- Sequential decoding
- Feedback Coding