

V sem

Unit - 4

Information Theory and Coding:- Information measurement,

Average information and information rate, coding for discrete memoryless source, continuous channel capacity

Maximum entropy, Huffman and Fano coding,

Discrete channel capacity, Trade-off between SN and Bandwidth, Block code, Hamming code,

cyclic code, Convolution Code, Tree diagram,

State diagram, Trellis diagram, Viterbi encoder and decoder, Turbo code.

Information Theory

→ The mathematical study of the coding of information in the form of sequence of symbols, impulses etc.

⇒ Information means importance.

✓ ⇒ if the probability of occurrence of an event is less then the information associated with that event will be more and vice-versa.

Ex:- if a dog bites a man, it's no news, but if a man bites a dog it's a news.
 ex. 1 - the first

$$I(x_i) \propto \frac{1}{P(x_i)}$$

$$I(x_i) = \log_b \left\{ \frac{1}{P(x_i)} \right\}$$

or

$$I(x_i) = -\log_b \{ P(x_i) \}$$

* Unit of $I(x_i)$ depends upon the base (b) chosen

b	units
2	bits binary
e	nat (natural unit)
10	decit or Hartley Hartley

⇒ it is standard to use $b=2$. the conversion of these units to other unit can be obtained by the following relations:

$$\log_2 a = \frac{\log_{10} a}{\log_{10} 2}$$

Ex-1 calculate the amount of information if it is given that $P(x_i) = \frac{1}{4}$.

Soln:- we know that the amount of information is given as

$$I(x_i) = \log_2 \left(\frac{1}{P(x_i)} \right) = \frac{\log_{10} \frac{1}{P(x_i)}}{\log_{10} 2}$$

Substituting given value of $P(x_i)$ in above eqn

$$I(x_i) = \frac{\log_{10} 4}{\log_{10} 2} = \underline{\underline{2 \text{ bits}}}$$

$$\frac{\log_{10} 4}{\log_{10} 2} = \frac{2 \cdot \log_{10} 2}{\log_{10} 2} = 2$$

Q.2 A source is generating 8 possible systems with probabilities of $\frac{1}{4}, \frac{1}{4}, \frac{1}{4}$. Find the information associated with each of the symbols.

Soln: Given $P(x_1) = \frac{1}{4}, P(x_2) = \frac{1}{4}, P(x_3) = \frac{1}{4}$

$$I(x_1) = \log_2 \left(\frac{1}{P(x_1)} \right) = \log_2 4 = 2 \text{ bits}$$

$$I(x_2) = \log_2 \left(\frac{1}{P(x_2)} \right) = \log_2 4 = 2 \text{ bits}$$

$$I(x_3) = \log_2 \left(\frac{1}{P(x_3)} \right) = \log_2 4 = 2 \text{ bits}$$

Note: the probability of occurrence of x_2 is high so that the information associated with x_2 will be less.

Q.3 Show that $I(x_1 x_2) = I(x_1) + I(x_2)$ if x_1 and x_2 are independent.

Soln: If x_1 and x_2 are independent then

$$P(x_1 x_2) = P(x_1) \cdot P(x_2)$$

$$I(x_1 x_2) = \log \left(\frac{1}{P(x_1 x_2)} \right)$$

$$= \log \frac{1}{P(x_1) \cdot P(x_2)}$$

$$= \log P(x_1) + \log \left(\frac{1}{P(x_2)} \right)$$

$$= I(x_1) + I(x_2)$$

Ex-4 Calculate the amount of information if binary digit (bits) occurs with equal likelihood in a binary PCM systems.

Soln: In binary PCM systems there are only two binary levels i.e. 1 or 0. Since they occur with equal likelihood, their probability of occurrence would be

$$P(x_1) = P(x_2) = \frac{1}{2}$$

Therefore the amount of information content will be given as

$$I(x_1) = \log_2 \left(\frac{1}{P(x_1)} \right) = \log_2 2 = 1$$

$$I(x_2) = \log_2 \left(\frac{1}{P(x_2)} \right) = \log_2 2 = 1$$

$$I(x_1) + I(x_2) = 1 + 1 = 2$$

Hence the correct identification of binary digit (bits) in binary PCM carries 1 bit of information.

Q.5 In a binary source, 0s occur three times as often as 1s. Find the information contained in 0s and 1s. Express the information in bits, nats, and Hartleys.

Soln: If prob. of symbol occurring in data is $P(0)$, and 1 for $P(1)$

$$P(0) + P(1) = 1$$

$$3P(0) + P(1) = 1$$

$$P(0) = \frac{3}{4}, P(1) = \frac{1}{4}$$

$$I(x_0) = \log_2 \left(\frac{1}{P(x_0)} \right) = -\log_2 P(x_0) = -\log_2 \frac{3}{4} = 0.585 \text{ bits}$$

$$I(x_1) = -\log_2 P(x_1) = -\log_2 \frac{1}{4} = 2 \text{ bits}$$

A key measure in information theory is "entropy".

Average Information (Entropy) :

In a practical communication system, we usually transmit long sequence of symbols from a source produces a series of symbols. Thus we are more interested in the average information than the information content of a single symbol.

Entropy quantifies the amount of uncertainty in the value of a random variable or the outcomes of a random process.

$$H = \sum_i I(x_i) \cdot P(x_i)$$

$$H = \sum_i P(x_i) \log_2 \left(\frac{1}{P(x_i)} \right)$$

$$H = - \sum_i P(x_i) \cdot \log_2 P(x_i)$$

* it is measure of the avg information content per source symbol.

Entropy is the avg information in the symbols produced by a msg source.

* Information Rate (R) : The avg rate at which the information must be transmitted is called the source of information rate.

⇒ Now,

$$R = \frac{\text{bits}}{\text{Symbol}} \times \frac{\text{Symbol}}{\text{Rate}}$$

$$R = H \times r$$

$$\Rightarrow \text{Information Rate} = \frac{\text{Symbol}}{\text{Rate}} \times \text{Entropy}$$

* bits stands for binary digits as message or code element that is different from information bit.

Q-1

A source is generating 4 possible symbols with the probability of $\frac{1}{8}, \frac{1}{8}, \frac{1}{4}, \frac{1}{2}$. Find the entropy and information rate if the source is generating 1 symbol/msec.

Given:- $P(x_1) = \frac{1}{8}, P(x_2) = \frac{1}{8}, P(x_3) = \frac{1}{4}, P(x_4) = \frac{1}{2}$

Symbol rate = 1 symbol/msec.

$$r = 1000 \text{ symbol/sec}$$

$$H = - \sum_{i=1}^4 P(x_i) \log_2 \left(\frac{1}{P(x_i)} \right)$$

$$= \frac{1}{8} \log_2 8 + \frac{1}{8} \log_2 8 + \frac{1}{4} \log_2 4 + \frac{1}{2} \log_2 2$$

$$H = 1.75 \text{ bits/symbol}$$

$$R = r \times H$$

$$= 1000 \times 1.75$$

$$R = 1.75 \text{ Kbps}$$

* Entropy is measure of uncertainty.

Analysis:



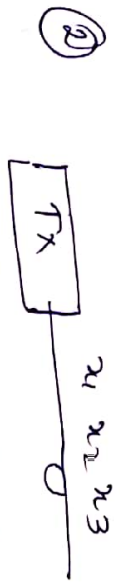
Case-1: $P(x_1) = P(x_2) = \frac{1}{2}$

$$H = -\sum_{i=1}^2 P(x_i) \log_2 P(x_i)$$

$$= -\frac{1}{2} \log_2 \frac{1}{2} - \frac{1}{2} \log_2 \frac{1}{2}$$

$$= 2 \text{ bits/symbol}$$

when the prob. are equal.



Case ①: $P(x_1) = P(x_2) = P(x_3) = \frac{1}{3}$

$$H = \log_2 3 \text{ bit/symbol}$$

when the probability of 1 bit and other is zero.

Case ②: $P(x_1) = 1, P(x_2) = P(x_3) = 0$



$$P(x_1) = P(x_2) = P(x_3) = \dots = P(x_m) = \frac{1}{m}$$

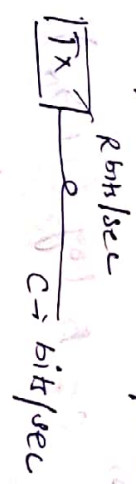
$$H_{max} = \log_2 m \text{ bits/symbol}$$

$$H_{min} = 0 \text{ bits/symbol}$$

channel capacity:

↳ it specifies the no of bits allowed by the channel in 1 sec.

* Channel Capacity, $C = \text{bits/sec}$



$$C \geq R \leftarrow \text{No information loss}$$

Shannon Hartley Law:

↳ It gives the relation b/w channel capacity (C) and Bandwidth (BW)

↳ Mathematically,

$$C = B \log_2 (1 + \frac{S}{N})$$

Normal SN (not in dB)

$$C = \text{channel capacity (bits/sec)}$$

$$B = \text{channel BW (Hz)}$$

$$\frac{S}{N} \text{ dB} = 10 \log_{10} (\frac{S}{N})$$

S = Channel power expected at channel output
 N = Noise power
 (avg signal power)

$\frac{S}{N}$ (dB)	$\frac{S}{N}$
10 dB	10
20 dB	100
15 dB	101.5 =

$10 \text{ dB} = 10 \log_{10} (\frac{S}{N})$
 $1 = \log_{10} (\frac{S}{N})$
 $10^1 = 10^1 = 10$

Q-1 For a channel of $B_{w} = 4 \text{ KHz}$

$(S/N) = 15 \text{ dB}$

Find the channel capacity

Sol: As $S/N = 15 \text{ dB}$

$= 10^{1.5} = 31.6$

So,

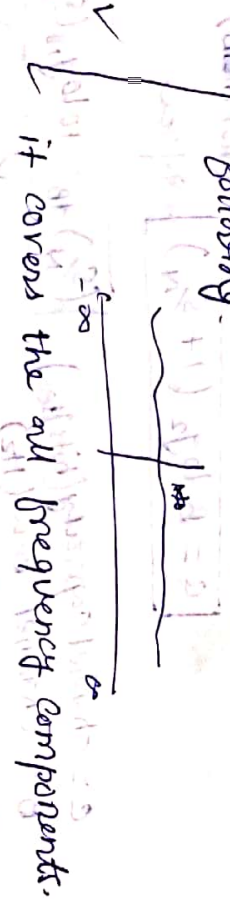
$C = B \log_2 (1 + S/N)$

$= 4 \log_2 (1 + 31.2)$

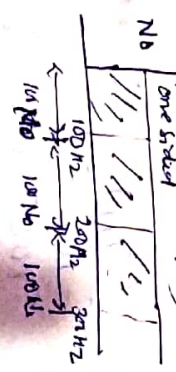
$C = 20.1 \text{ Kbps}$

~~Channel~~ Capacity of AWGN (Additive white Gaussian Noise) Channel is -

White noise has the frequency spectrum as following.



→ The PSD of the white noise is given as



* freq. b/w 0 to 100Hz will be affected by 100Watts of power.

* freq. b/w 100 to 300Hz will be affected by 100Watts of power.

if the line freq. is also considered then

PSD of below only $N_0/2$

Regarding white noise, its power is given as

$N \text{ (Watts)} = \frac{\text{Watts}}{\text{Hz}} \times \text{Hz}$

$N = N_0 B$

* Default power spectral density is one sided.

Note:- Each of the frequency component transmitted through a channel is affected by same amount of white noise power.

The channel B_w is given as

$C = B \log_2 (1 + \frac{S}{N})$
(Linear)

For a AWGN channel,

$C = B \log_2 (1 + \frac{S}{N_0 B})$
(nonlinear)

Conclusion:-

For AWGN channel as $B \rightarrow \infty$
channel capacity becomes $C_{\infty} = 1.44 \frac{S}{N_0}$

Prob: As we know that

$$C = B \log_2 \left\{ 1 + \frac{S}{N_0 B} \right\}$$

$$C = \frac{N_0}{S} \cdot \left(\frac{S}{N_0} \right) B \log_2 \left\{ 1 + \frac{S}{N_0 B} \right\}$$

Let $\frac{N_0 B}{S} = x$

As $B \rightarrow \infty \Rightarrow x \rightarrow \infty$

$$C_{\infty} = \lim_{x \rightarrow \infty} \frac{S}{N_0} \cdot x \cdot \log_2 \left(1 + \frac{1}{x} \right)$$

$$= \frac{S}{N_0} \log_e e$$

$$C_{\infty} = 1.44 \frac{S}{N_0}$$

Bandwidth SN Trade off:-

For a noiseless channel ($N_0 = \infty$) has finite capacity. On the other hand,

the channel capacity does not become infinite as the B becomes approaches infinite because with an increase of B , the noise power also increases.

Thus for a fixed signal power, and in the presence of white gaussian noise, the channel capacity approaches an upper limit with increasing B . This is explained below. We have,

$$C = B \log_2 \left(1 + \frac{S}{N_0 B} \right)$$

$$= \frac{S}{N_0} \cdot \frac{N_0 B}{S} \cdot B \log_2 \left(1 + \frac{S}{N_0 B} \right)$$

$$= \frac{S}{N_0} \cdot \frac{N_0 B}{S} \log_2 \left(1 + \frac{S}{N_0 B} \right)$$

Let $\frac{N_0 B}{S} = x$

$$= \frac{S}{N_0} \cdot x \cdot \log_2 \left(1 + \frac{1}{x} \right)$$

$$= \frac{S}{N_0} \cdot \lim_{x \rightarrow \infty} \left(1 + \frac{1}{x} \right)^{1/2} = e$$

As B approaches infinite, $\frac{N_0 B}{S}$ approaches zero.

$$\lim_{B \rightarrow \infty} \left(1 + \frac{S}{N_0 B} \right)^{N_0 B/S} = e$$

Hence,

$$\lim_{B \rightarrow \infty} C = \frac{S}{N_0} \log_e e$$

$$C_{\infty} = 1.44 \frac{S}{N_0}$$

Let us now consider the tradeoff b/w the B.W and S/N ratio

$$\left(\frac{S}{N} \right) = 15, B = 5 \text{ KHz}$$

$$\left(\frac{S}{N} \right) = 31, B = 20 \text{ KHz}$$

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

$$= 5 \log_2 (1 + 15)$$

$$= 5 \log_2 16 \Rightarrow 5 \log_{10} 16$$

$$= \frac{5 \times 1.204}{\log_{10} 2}$$

$$= 5 \times 4 = 20 \text{ Kbps}$$

If the S/N ratio is increases to 31, the B.W for the same channel capacity can be found in form

$$C = 20 = B \log_2 (1 + 31)$$

$$B = \frac{20}{\log_2 32} = 4 \text{ KHz}$$

with a 4 KHz B.W, the noise power will be $\frac{1}{5}$ times the noise power at 5 KHz, Thus the signal power will have to increase by factor $\frac{1}{5} \times \frac{31}{15} = 1.65$.

Q-1 For AWGN of having B.W 4 KHz, two sided noise PSD is given by 10^{-12} W/Hz. Find the channel capacity required to get signal power of 0.1 mW at the output of the channel.

Given:- B.W = 4 KHz

$$N_0 = 10^{-12}$$

$$N_0 B = 2 \times 10^{-12}$$

$$N_0 B = 2 \times 10^{-12} \times 4 \times 10^3$$

$$= 8 \times 10^{-9} \text{ W/Hz}$$

$$C = B \log_2 \left(1 + \frac{S}{N_0 B} \right)$$

$$= 4 \times 10^3 \log_2 \left(1 + \frac{0.1 \times 10^{-3}}{8 \times 10^{-9}} \right)$$

$$C = 84.44 \text{ kbps}$$

18.2

A voice grade channel for the telephone channel has BOD of 7.4 kHz. Calculate the information capacity of the telephone channel for signal to noise ratio of 30 dB

Soln:- Given $\frac{S}{N} / dB = 30 dB$

$$\left(\frac{S}{N}\right)_{dB} = 30 dB = 10 \log \left(\frac{S}{N}\right)$$

$$30 = 10 \log \left(\frac{S}{N}\right)$$

$$\frac{S}{N} = 10^3 = 1000$$

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

$$= 3.4 \times 10^3 \log_2 (1 + 1000)$$

$$= 3.4 \times 10^3 \log_2 (1001)$$

$$C = 33.8 \times 10^3 \text{ bps}$$

coding offers most significant application of the information theory. The main purpose of coding is to improve the efficiency of the communication system in some sense.

Entropy coding:-

⇒ The design of a variable length code such that its average code word length approximates the entropy of the DMS is referred to as entropy coding. In this coding section we presents two examples of entropy coding;

(1) Huffman encoding

(11) Shannon-Fano coding

(i) Huffman encoding:-

We describe an important class of prefix codes known as Huffman codes.

✓ In general, Huffman encoding results in an optimum code. Thus, it is the code that has the highest efficiency.

⇒ We present the procedure using a specific example.

Suppose that we wish to code five words x_1, x_2, x_3, x_4 and x_5 with probabilities $P(x_1) = 0.4, P(x_2) = 0.19, P(x_3) = 0.16, P(x_4) = 0.15, P(x_5) = 0.1$. The Huffman code is accomplished in four steps as follows.

STEP-1

* Arrange the messages in order of decreasing probabilities.

* If there are equal probabilities choose any of the various ordering.

* The two source symbols of lowest probability are assigned a '0' and a '1'.

Words	Probability
x ₁	0.4
x ₂	0.19
x ₃	0.16
x ₄	0.15
x ₅	0.1

Step 2

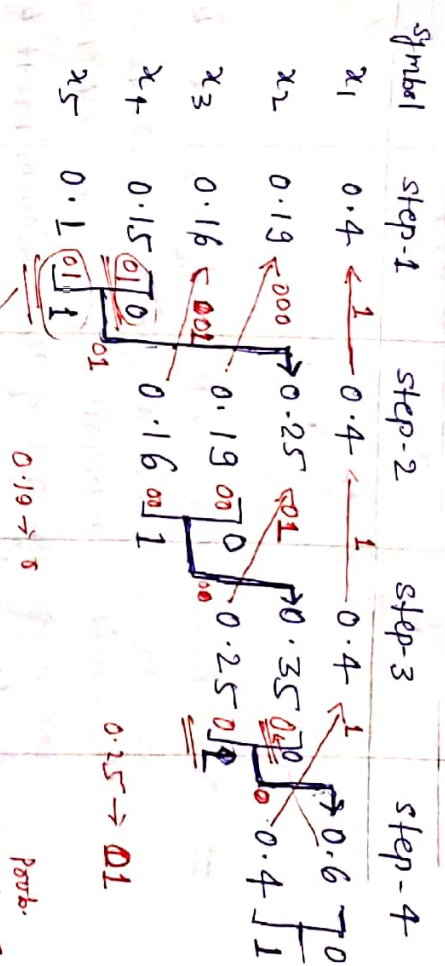
⇒ Combine the bottom two entries to form a new entry with a probability that is the sum of original probability.
 ⇒ if necessary, reorder the list so that new probabilities are in decreasing order.

Word	Probability (step-1)	Probability (step-II)
x ₁	0.4	0.4
x ₂	0.19	0.25
x ₃	0.16	0.19
x ₄	0.15	0.16
x ₅	0.1	0.1

Step-3

⇒ Continue combining in pairs until only two entries remain.

⇒ The codeword for each symbol is found by starting at the right with the most significant bit working backward and tracking the sequence of 0s and 1s assigned to that symbol as well as its successors.



Finally, the code words of the Huffman code for the source are tabulated in table (below)

Symbol	Probability	Code word
x ₁	0.4	1
x ₂	0.19	000
x ₃	0.16	001
x ₄	0.15	010
x ₅	0.1	011

The entropy is given by $H(X) = - \sum_{i=1}^n P(x_i) \log_2 P(x_i)$

$$= -0.4 \log_2 0.4 - 0.19 \log_2 0.19 - 0.16 \log_2 0.16 - 0.15 \log_2 0.15 - 0.1 \log_2 0.1$$

$$H(X) = 2.15$$

Average length is given by $L = 0.4 \times (1) + 0.19 \times (3) + 0.16 \times (3) + 0.15 \times (3) + 0.1 \times (3) = 2.2$

L = letters/message

$$\eta = \frac{H(x)}{L} = \frac{2.15}{2.2} = 0.977$$

$$\boxed{\eta = 97.7\%}$$

⇒ One disadvantage of the Huffman code is that we can not start assigning code words until the entire combination process is completed. That is, every one of the columns must be developed before the first code can be assigned.

* The coding process is often performed by a special purpose micro computer.

Shannon-Fano code:

✓ ⇒ Shannon-Fano code is similar to the Huffman code, with a major difference being that the operations are being performed in a forward, rather than backward, direction.

↳ Thus the storage requirements are considerably relaxed and the code is easier to implement.

⇒ While it often leads to average lengths that are the same as those of the Huffman code, the results of Shannon-Fano coding are not always good as those of Huffman coding.

⇒ We again illustrate the techniques with an example of that of Huffman codes presented earlier.

Step-1 Arrange the message in order of decreasing probability. If there are equal probabilities choose any of the various orderings.

Word	Probability
x_1	0.4
x_2	0.19
x_3	0.16
x_4	0.15
x_5	0.1

Step-2

⇒ Partition the message into the most equiprobable subsets that is start at the top or bottom and divide the group into two sets.

⇒ We find the total probability of upper set and ~~total~~ the total probability of the lower set.

⇒ we choose the dividing line that results in the closest two probability.

⇒ we now assign a 0 to all members of one of the two sets and a 1 to all members of the others (the choice is arbitrary).

⇒ suppose we choose 0 for the top set and 1 for the bottom. this is illustrated here.

Words	Probability	STEP-1
x_1	0.4	0
x_2	0.19	0
x_3	0.16	1
x_4	0.15	1
x_5	0.1	1

ie it divided in two parts, 0.59 and 0.41 almost similar

Step-3

Continue this process, each time partitioning the sets with as nearly equal probability as possible until further partitioning is not possible.

Finally Shannon-Fano code is constructed as follows.

Symbols	Probability	step-1	step-2	step-3	code
x_1	0.4	0	0		00
x_2	0.19	0	1		01
x_3	0.16	1	0		10
x_4	0.15	1	1	0	110
x_5	0.1	1	1	1	111

This entropy is given by

$$H(x) = -\sum_{i=1}^5 P(x_i) \log_2 P(x_i)$$

$$= -0.4 \log_2 0.4 - 0.19 \log_2 0.19 - 0.16 \log_2 0.16 - 0.15 \log_2 0.15 - 0.1 \log_2 0.1$$

$$H(x) = 2.18$$

Average length is given by

$$L = \sum_{i=1}^5 P(x_i) \cdot n_i$$

$$= 0.4(2) + 0.19(2) + 0.16(2) + 0.15(3) + 0.1(3)$$

$$= 2.25$$

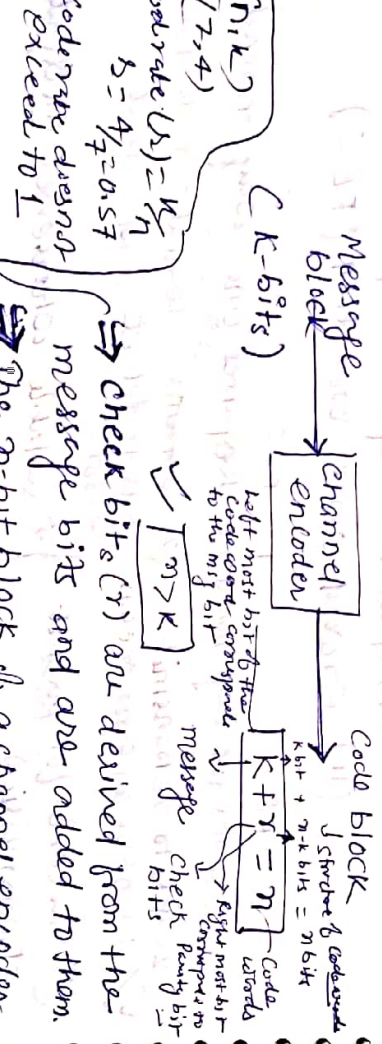
$$\eta = \frac{H(x)}{L} = \frac{2.18}{2.25} = 0.968 = 96.8\%$$

Error control coding is classified in two types

- ① Block Code
- ② Convolution code.

⇒ Block Code: In the

⇒ Block code is a code having all its words of the same length,
 ⇒ k bit block is converted into an n-bit block, or each block of k message bits is encoded into a block of n bits (n > k).
 ⇒ The resultant block code called an (n, k) block code.



⇒ check bits (r) are derived from the message bits and are added to them.
 ⇒ The n-bit block of a channel encoder output is called a code word.
 * Parity bits help in error detection and correction, also in locating data, i.e. in the data being transmitted

② Parity check code:

⇒ The simplest possible block code is when the no of check bits is one. There are known as parity check code.
 ⇒ when total no of 1s in the code word is Even then this is called even parity check code.

⇒ and when the total no of 1s in the code word is odd, then this is called odd parity check code.

Following example explains the parity check code

Message

010011 [no. of 1's odd]	Code for even parity (check bit)	Code for odd parity (check bit)
101110 [no. of 1's even]	0	1

③ Binary code space:

→ In this section certain important concepts such that as the weight code, Hamming distance etc.

Hamming weight :-
 ⇒ The weight of the code = no of non-zero components

Code words	Weight
010110	3
101000	2
000000	0

④ Hamming distance:

⇒ no of components in which two code words differs.

EX

U = 10110
 V = 01111
 W = 10011

Per form XOR operation

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

⇒ H = 3

cyclic code:-

$C = (000, 101, 011, 110)$
 $0000, 0101, 1010, 1111$

* e is 0 linear code.
 A code is said to be cyclic code if the cyclic shift of each code word is also a code word. i.e. whenever $a_0, \dots, a_{n-1} \in C$, then also $a_{n-1}, a_0, \dots, a_{n-2} \in C$.

Theorem

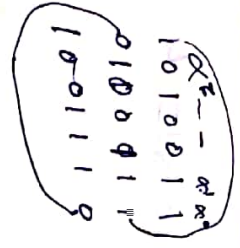
\Rightarrow If $g(x)$ is a polynomial of degree $(n-k)$ that is a factor of $(x^n - 1)$, then $g(x)$ generates an (n, k) cyclic code C in which the code polynomials $c(x)$ for the data word d are generated by

$$c(x) = d(x) \cdot g(x)$$

where $d(x) = d_0 + d_1x + d_2x^2 + \dots + d_{k-1}x^{k-1}$ is the data polynomial corresponding to the data word.

\therefore Polynomial $g(x)$ is called generator polynomial.

Ex:



$$C(x) = x^6 + x^4 + x^2 + 1$$

Polynomial Representation of code word

Generator Matrix:-

\Rightarrow In an (n, k) linear block code C , we represent a code vector c and data vector d as follows.

$$c = [c_1, c_2, c_3, \dots, c_n]$$

$$d = [d_1, d_2, d_3, \dots, d_n]$$

\Rightarrow If a data bit appear in specified location of c , then the code c is called systematic otherwise non-systematic.

$$c_1 = d_1$$

$$c_2 = d_2$$

$$c_3 = d_3$$

$$c_n = d_n$$

$$C = dG$$

$$C = [d_1, d_2, d_3, \dots, d_n]$$

$$G = \begin{bmatrix} 1 & 0 & 0 & 0 & \dots & 0 & p_{11} & p_{12} & \dots & p_{1n} \\ 0 & 1 & 0 & 0 & \dots & 0 & p_{21} & p_{22} & \dots & p_{2n} \\ 0 & 0 & 1 & 0 & \dots & 0 & p_{31} & p_{32} & \dots & p_{3n} \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 1 & p_{n1} & p_{n2} & \dots & p_{nn} \end{bmatrix}$$

We know that, $c = dG$, then

$$G = [I_k \quad P^T]$$

$P^T =$ Transpose of P

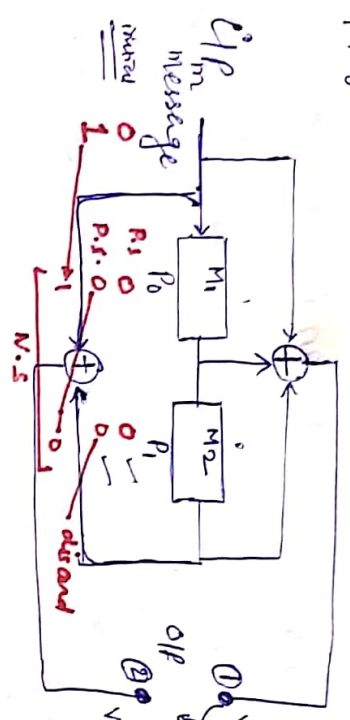
$I_k =$ Identity matrix

x_{0-2}
 $0\ 0\ 0$
 $0\ 1\ 1$
 $1\ 1\ 1$
 $1\ 1\ 0$

1^{st}
 $100 \oplus 0 = 111$
 $100 = 111$

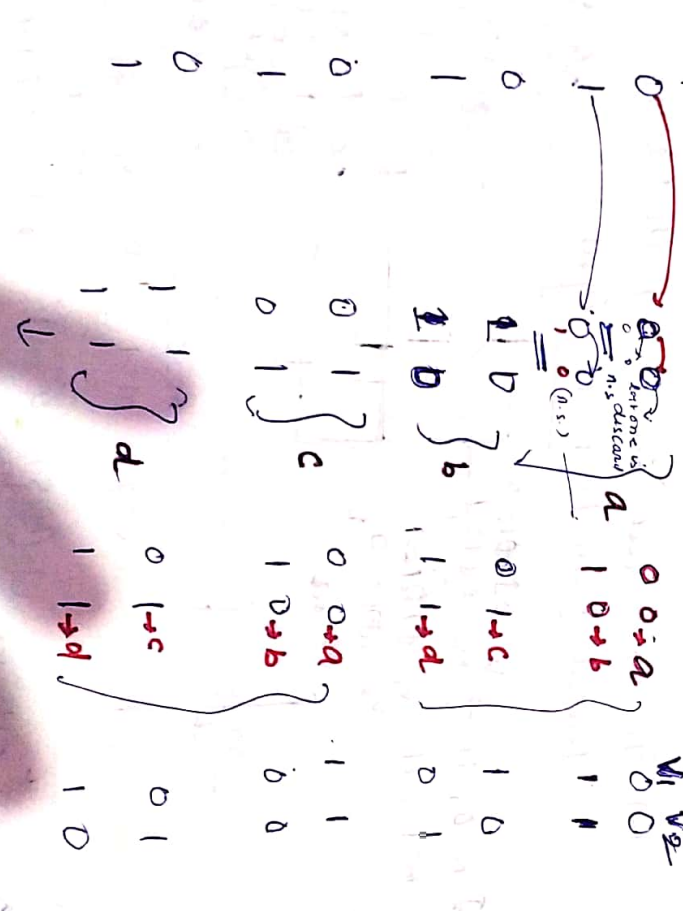
Convolution Code

Encoder Convolution
 1^{st}
 $00 \rightarrow R$
 $10 \rightarrow b$
 $01 \rightarrow c$
 $11 \rightarrow d$
 $00 \oplus M_1 \oplus M_2$
 $10 \oplus M_1 \oplus M_2$
 $01 \oplus M_1 \oplus M_2$
 $11 \oplus M_1 \oplus M_2$



$M_1, M_2 \rightarrow$ Memory [Actually shows Present state (P.S.)
 a, P_0, P_1 here two memory, so that gives combination $00, 01, 10, 11$

Next state (N.S.)
 Present state (P.S.)
 P_0, P_1

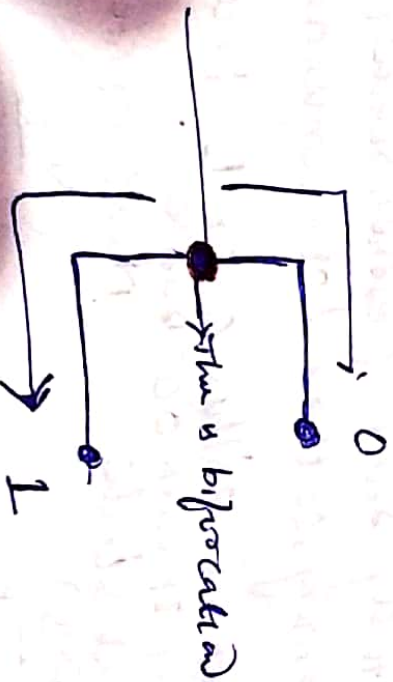


In case of N.S. - every bit is shifted

Input is shifted in the first memory element
 and first memory element is shifted to the 2nd memory element then finally get N.S.

Tree diagram:

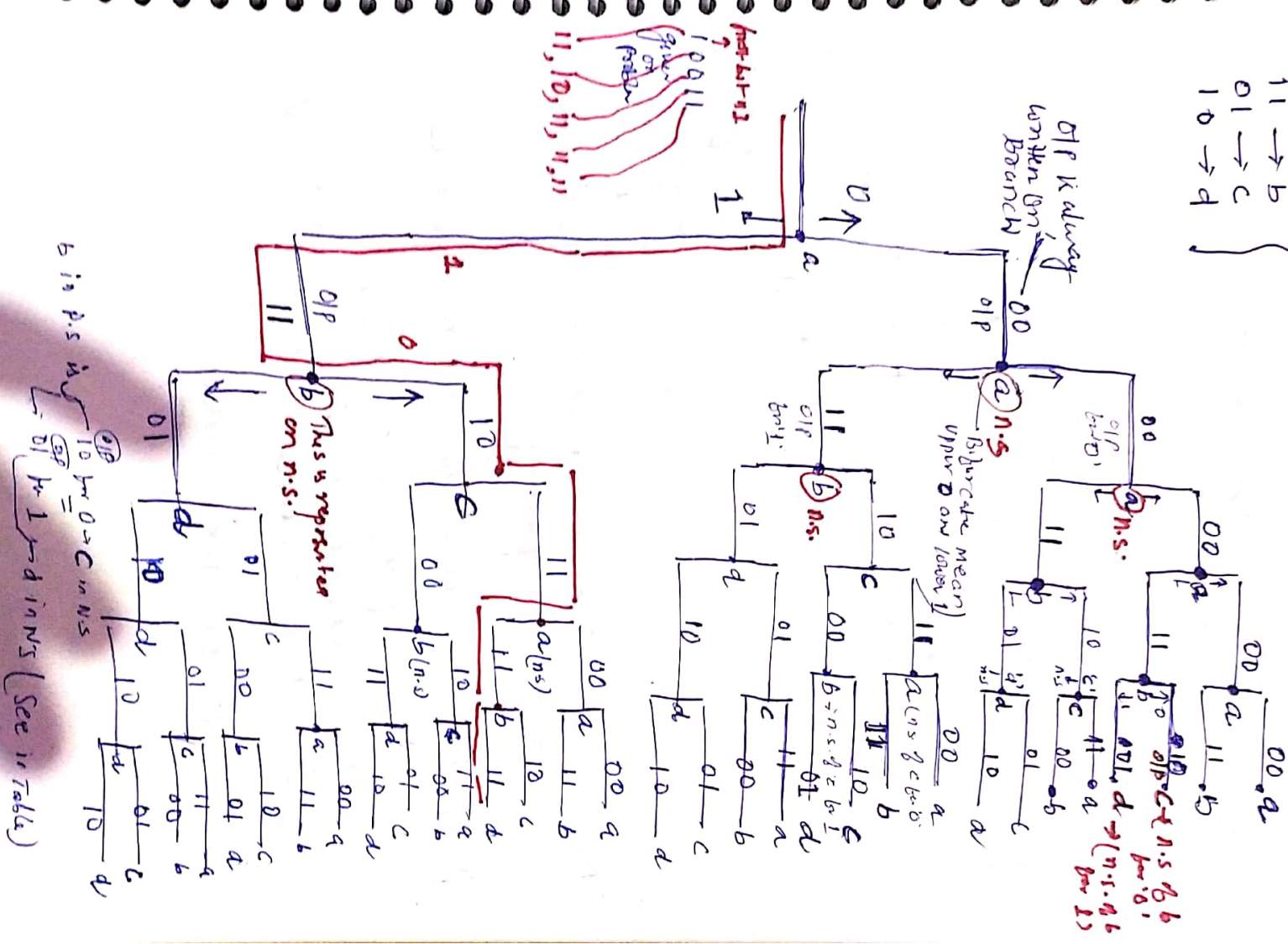
- (i) Every branch represents an input bit
- (ii) The upper branch of the bifurcation (Division of a branch) denotes a bit '0'
- (iii) The lower branch of the bifurcation denotes a bit '1'



- (iv) The enclosed sequence path has to be traced from left to right —
M₁ M₂

00 → a
 11 → b
 01 → c
 10 → d

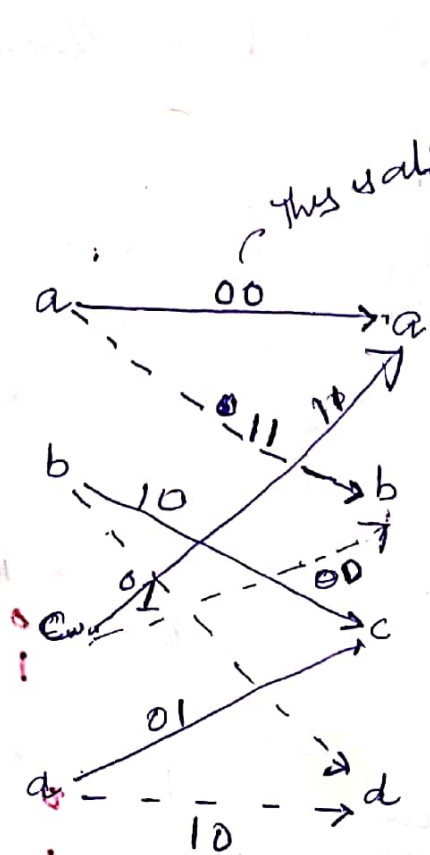
10011
 gives
 feedback
 11, 10, 11, 11



b is p.s in 10 for 0 → C in n.s.
 DIP for 1 → d in n.s. (See in Table)

Trellis Diagram :-

a → for 0 → ———→ solid line
 for 1 → - - - - -→ dash line



Case (1) when 'a' then it moving towards a + b [from N.S] off (00, 11)

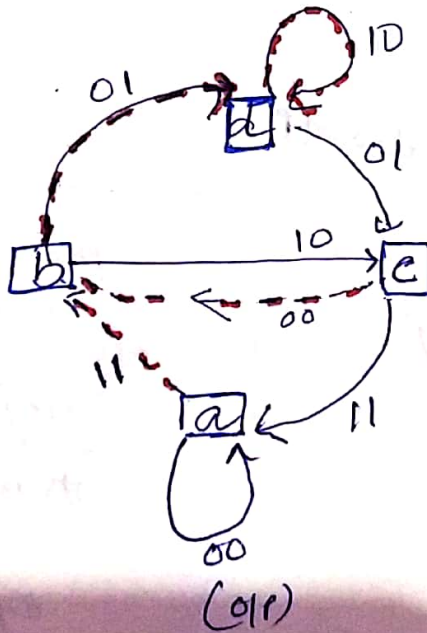
Case (2) when b then its moving towards e + d (N.S) 01, 10 11, 01

Case (3) when c then its moving towards a and b 01, 11, 00

Case (4) d then c, d 01, 10

State Diagram :- There are four state, a, b, c, d

0 → Solid line
1 → dash line



	N.S	o/p
a → a	00	00
a → b	01	11

Continue Cyclic code

See Previous Page explanation

In code word (7, 4)

1110 110

↳ not differentiate msg bit and parity check bit in case
 ↳ if given code word we don't know the arrangement of which side is msg and which side is parity (or mixed) then this type of code words is known as Non-Systematic Code word.

↳ we can identify msg bit and parity check bit if arrangement of given code words is known which first four bit is msg and last 3 bit parity or first 3 bit is msg or last four bit msg. this type of code is known as systematic Code

This type of code is generally found in cyclic code.

Example of systematic code is "linear Block Code"

Ex. of non-systematic codes
 Convolutional code. Cyclic code word

Non-systematic Code word

$$X(x) = M(x) \cdot g(x)$$

Code word
Polynomial of msg
↳ generator Polynomial

↳ both are given generally
 ↳ if not given then we assume msg then solve.

Systematic Code word

$$\textcircled{1} X(x) = x^{n-k} M(x)$$

$$\textcircled{2} \frac{x^{n-k} M(x)}{G(x)} \rightarrow \text{Quotient} + \text{Remainder}$$

$$\textcircled{3} X(x) = \frac{x^{n-k} M(x)}{G(x)} + R(x)$$

Remainder

Q.2

Construct a Non-systematic and systematic (7, 4) cyclic code using a generator polynomial

$$g(x) = x^3 + x^2 + 1$$

Soln!

$$C(x) = M(x) \cdot g(x)$$

let $M(x) = 1010$

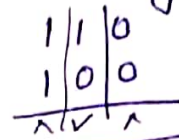
$$= 1 \cdot x^3 + 0 \cdot x^2 + 1x + 0 \cdot x^0$$

$$= x^3 + x$$

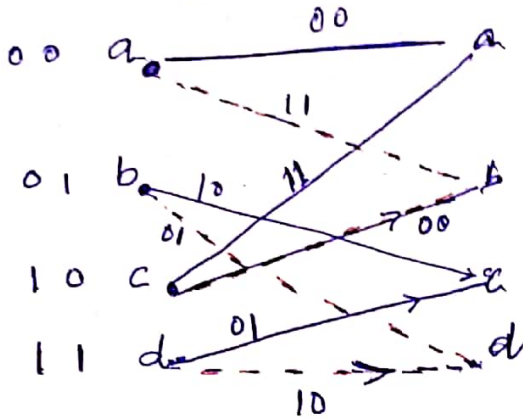
Viterbi Algorithm (VA) for Decoding of Convolutional codes:

*The Viterbi Algorithm is used to decode ~~the~~ convolutional codes and any structure or system that can be described by a trellis.

✓ Before starting, we should have idea about Hamming distance
suppose,



Hamming distance = 1

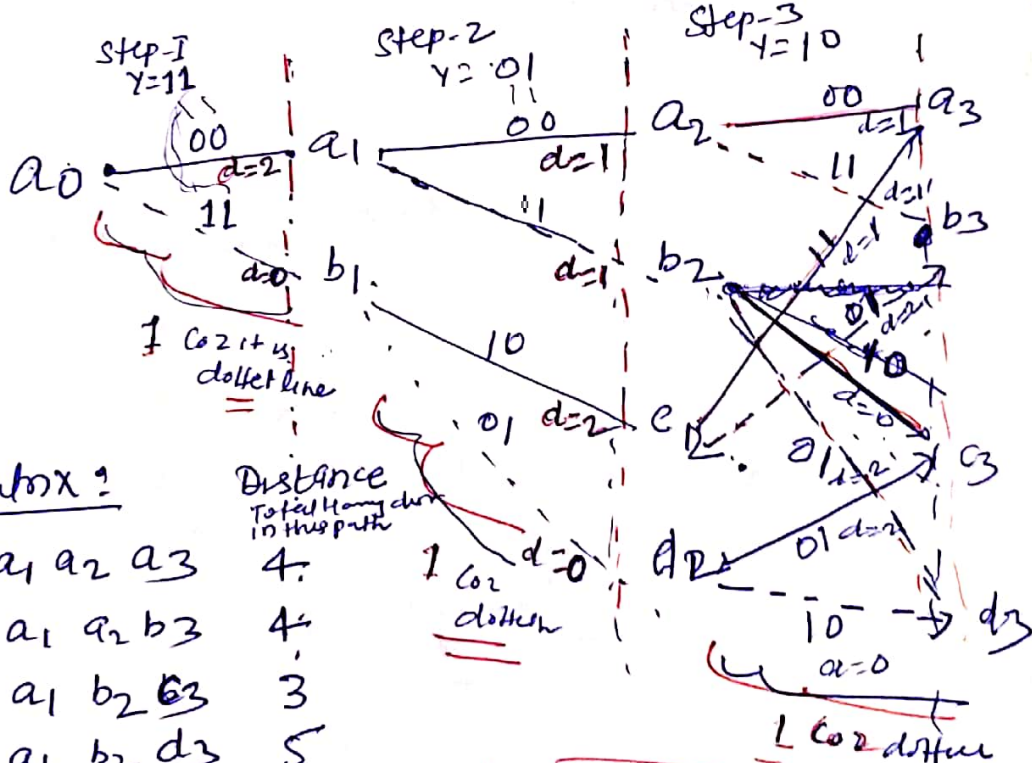


So on we count the total no of errors on each possible path and choose the path with the fewest errors. we do this one step at a time

** Viterbi decoding tells us to choose the smallest number at each state and eliminate the other path.

↳ Suppose we have received a input, and that input we want to decode.

↳ we receive input $Y = 110110$ → code word then how to decode this



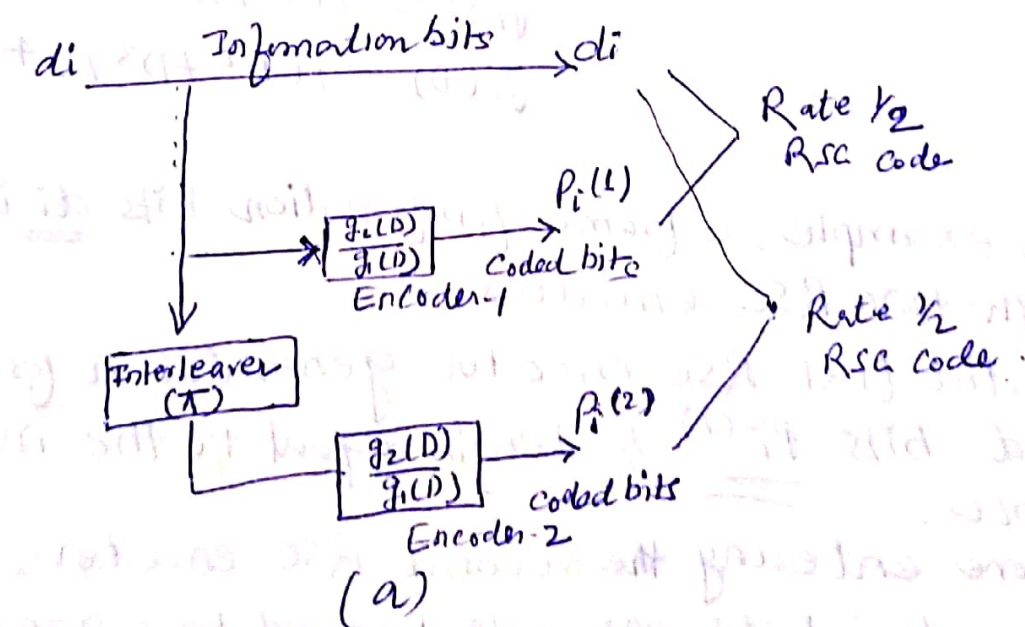
Path Matrix:

1.	a_0	a_1	a_2	a_3	4.
2.	a_0	a_1	a_2	b_3	4.
3.	a_0	a_1	b_2	c_3	3
4.	a_0	a_1	b_2	d_3	5
5.	a_0	b_1	c_2	a_3	3
6.	a_0	b_1	c_2	b_3	3
7.	a_0	b_1	d_2	c_3	2
8.	a_0	b_1	d_2	d_3	0

∴ $X = 111$ encoded
We choose the path with the smallest sum. The information sequence (ie input to the encoder) 111 corresponds to path
→ if we get '0' then that is the best path, then in Receive Code there is no error

Turbo Codes :-

- ↳ Turbo codes are combination of Block and Convolutional Codes.
- ↳ They require the a block to be formed before encoding can begin, but for encoding shift register are used, Just like in convolutional code instead of computing parity bits.
- ↳ Turbo codes are used at least two convolutional components encoders and two maximum a posteriori (MAP) algorithm components decoders.
- ✓ ⇒ Turbo codes perform well in the low SNR environment.
- ✓ ⇒ At high SNR, Reed-Solomon code have better performance than Turbo codes.



⇒ This turbo consists of two recursive systematic convolutional (RSC) codes.

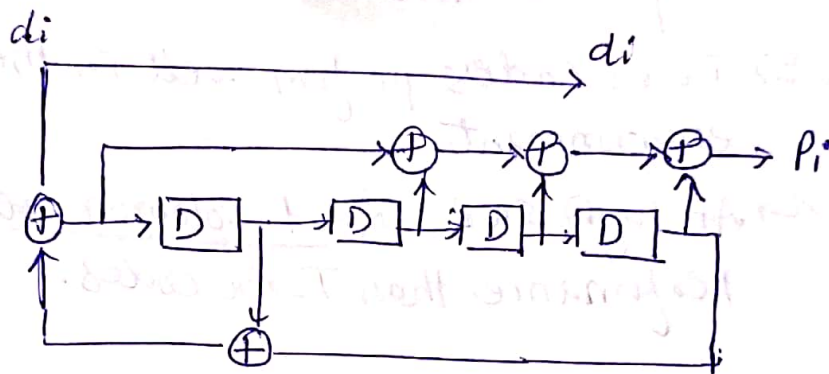
↳ Unit delay as D , the 1×2 generator matrix of the rate $1/2$ RSC code is of the form.

$$G(D) = \begin{bmatrix} 1 & \frac{g_2(D)}{g_1(D)} \end{bmatrix}$$

In particular, the example turbo codes of Berrou et al. was specified by

$$G(D) = \begin{bmatrix} 1 & \frac{1+D^2+D^3+D^4}{1+D+D^4} \end{bmatrix}$$

The simple implementation of encoder is shown figure below.



$$g_1(D) = 1 + D + D^4$$

$$g_2(D) = 1 + D^2 + D^3 + D^4$$

⇒ In this example, a frame of information bits $\underline{d_i}$ is sent through two RSC encoders.

Thus, the first RSC encoder generates a frame of coded bits $\underline{p_i^{(1)}}$ of length equal to the information frame.

↳ Before entering the second RSC encoder, the information bits are interleaved by a random block interleaver π . The second encoder will generate a different coded bit frame $\underline{p_i^{(2)}}$