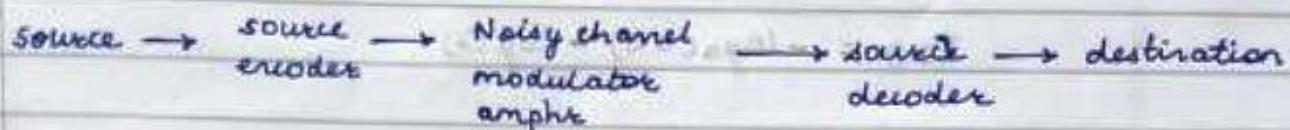


Information theory and coding

- Information \rightarrow message

Information can be analog or discrete



- Measure of Information

Information is transmitted from source which is digital or analog. Let the discrete source descending symbols s_1, s_2 with prob. of occurrence p_1, p_2 . Out of these symbols if we choose s_k symbol with prob. p_k .

When prob. of occurrence is low that is uncertainty is high then it contains information, whereas, if uncertainty is high then there is less amount of information. So the measure of inf. was quantitatively defined by shannon

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

\downarrow

symbol s_k 's probability p_k

I_k is self information of message s_k and b is base of logarithm. Since $0 \leq p_k \leq 1$, $I \geq 0$. Information is always non-negative

The amt of information in a message depends only on its probability of occ. rather than its actual content or its physical interpretation

Consider example of tossing a fair coin. A binary source is producing outcome 0 for tail and 1 for head

$$P(1) = 0.5, P(0) = 0.5$$

$$I(H) = -\log_2(0.5) = \log_2(2)$$

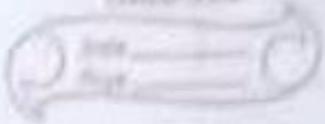
$$= \frac{-\log(0.5)}{\log(2)}$$

$$= 1 \text{ bit}$$

for a discrete memoryless source, each message is independent of previous message

Properties of Information

- i) More uncertainty of message, more the information required
- ii) If receiver knows message transmitted, the information 0
- iii) Total information for n symbols $s_0, s_1, s_2, \dots, s_n$ with prob. p_1, p_2, \dots
then total inf. $\rightarrow I_T = \log_2\left(\frac{1}{p_0}\right) + \log_2\left(\frac{1}{p_1}\right) + \log_2\left(\frac{1}{p_2}\right) + \dots$
- iv) If there are $M = 2^n$ equally likely messages, then amt of information carried by each message will be N bits
ex- the binary symbols 0 and 1 are transmitted with probabilities $P(0) = 1/4, P(1) = 3/4$ find the corresponding self information



$$I_0 = \log_2(4) \\ = 2 \text{ bits}$$

$$I_1 = \frac{\log 4}{\log 2} \rightarrow \log_2(4) = \frac{0.125}{0.301} \\ = 0.415 \text{ bits}$$

Highly probable unit gives less information

- Q-2- Prove the following statement - If the receiver knows the message transmitted, the amt of information is 0
 $P(k) = 1$
- Entropy:-

$$I_{\text{Total}} = L \left\{ P_0 \log_2 \left(\frac{1}{P_0} \right) + P_1 \log_2 \left(\frac{1}{P_1} \right) + P_2 \log_2 \left(\frac{1}{P_2} \right) \right\}$$

$$\underline{I_{\text{Total}}} = \left\{ P_0 \log_2 \left(\frac{1}{P_0} \right) + P_1 \log_2 \left(\frac{1}{P_1} \right) + P_2 \log_2 \left(\frac{1}{P_2} \right) \right\}$$

$H(S)$ - entropy = avg information

- Q- A discrete memoryless source produces 4 symbols n_1, n_2, n_3, n_4 with probability $P(n_1) = 0.5, P(n_2) = 0.2, P(n_3) = 0.2, P(n_4) = 0.1$
- $H(n)$. Find entropy or average information
Also obtain information contained $X_4 X_3 X_2 X_1$

$$\rightarrow \text{entropy} = \sum_{i=1}^4 P(n_i) \log_2 \left(\frac{1}{P_{X_i}} \right)$$

$$P_{n_1} \log_2 \left(\frac{1}{P_{n_1}} \right) + P_{n_2} \log_2 \left(\frac{1}{P_{n_2}} \right) + P_{n_3} \log_2 \left(\frac{1}{P_{n_3}} \right) + P_{n_4} \log_2 \left(\frac{1}{P_{n_4}} \right)$$

$$0.5 \log_2 \left(\frac{1}{0.5} \right) + 0.2 \log_2 \left(\frac{1}{0.2} \right) + 0.2 \log_2 \left(\frac{1}{0.2} \right) + 0.1 \log_2 \left(\frac{1}{0.1} \right)$$

$$= 0.5 + 0.46 + 0.464 + 0.332$$

$$H(x) = 1.6 \text{ bits/symbol}$$

b) $P(n_4 n_3 n_1 x_1) = 0.1 \times 0.2 \times (0.5)^2$
 $= 5 \times 10^{-3}$

$$I(n_4 n_3 n_1 x_1) = \log_2 \left(\frac{1}{P(n_4 n_3 n_1 x_1)} \right) = -\log_2 (5 \times 10^{-3})$$

$$= 7.643 \text{ bits/symbol}$$

a. Find the relationship between Hartley, nats, bits

$$\rightarrow I = \log_{10} \left(\frac{1}{P} \right) \text{ Hartley}$$

$$I = \log_e \left(\frac{1}{P} \right) \text{ nat}$$

$$I = \log_2 \left(\frac{1}{P} \right) \text{ bit}$$

$$1 H = I$$

$$\frac{\log_{10}(\frac{1}{P})}{\log_{10}(e)}$$

$$\left[\log_{10} b = \frac{1}{\log_b e} \right]$$

$$(H = \frac{\log_e(\frac{1}{P})}{\log_{10}(e)} \text{ nats})$$

$$\log_{10}(\frac{1}{P})$$

$$\frac{\log_e 10}{\log_e 2} = 2.303 \text{ nats}$$

$$\cdot 1 H = \frac{\log_2(\frac{1}{P})}{\log_{10}(e)} \text{ bits}$$

$$\frac{\log_e 10}{\log_e 2} = 3.32 \text{ bits}$$

$$\cdot 1 \text{ nat} = \log_2 e \text{ bits}$$

$$1 \text{ nat} = 1.443 \text{ bits}$$

e An analog

• Information Rate:-

let us suppose that symbols are emitted by source at a fixed time rate "ns" symbol/sec. The avg. source rate "Rs" in bits/sec are defined as product of (avg-information content per symbol) \times (message symbol rate, "ns")

$$R_s = H(s) \times r_s \text{ bits/sec}$$

$$R_s = \frac{H}{T}$$

where T is time required to send single message

Ex- An event has 4 possible outcomes with prob of occurrence $P_1 = \frac{1}{4}$, $P_2 = \frac{1}{2}$, $P_3 = \frac{1}{8}$, $P_4 = \frac{1}{8}$. Determine

the entropy of the system. Also obtain the rate of information if there are 8 outcomes per second

$$\rightarrow H = \frac{1}{4} \log 4 + \frac{1}{2} \log 2 + \frac{1}{8} \log 8$$

$$\frac{1}{2} + \frac{3}{8} + \frac{1}{2} + \frac{3}{8}$$

$$= 1 + \frac{3}{4}$$

$$= \frac{7}{4}$$

$$= 1.75 \text{ bits/message}$$

$$R_s = H(s) \times r_s$$

$$1.75 \text{ bits/message} \times 8 \text{ outcome/sec}$$

$$= 14 \text{ bits/sec}$$

- An output of information source consist of 150 symbol, 32 of which occur with prob. of $\frac{1}{64}$ and remaining 118 occur with prob of $\frac{1}{2^{36}}$. The source emits 2000 symbols per sec. assuming that symbols are chosen independently. Find avg information rate for the source

$$\rightarrow L = 150$$

$$H(x) = \sum_{i=1}^{32} P_{xi} \log_2 \left(\frac{1}{P_{xi}} \right) + \sum_{i=33}^{150} P_{xi} \log_2 \left(\frac{1}{P_{xi}} \right)$$

$$= 3 + 3.941$$

$$= 6.94 \text{ bits/message}$$

↓
symbol

$$R = H \times r$$

$$= 6.94 \times 2000$$

$$= 13880 \text{ bits/sec}$$

↓
symbol

$$= 13880 \text{ bits/symbol}$$

- Let us take first letter A to H, D fixed code length

letters	code used	letters	code used
A	000	E	100
B	001	F	101
C	010	G	110
D	011	H	111

variable code length (VLC) 1

Letters code used

A	00
B	010
C	011
D	100
E	101
F	110
G	1110
H	1111

Variable code length (VLC) 2

letter code used

A	0
B	1
C	00
D	01
E	10
F	11
G	000
H	111

Encode series of letters "ABAD CAB"

Message - ABAD CAB

fixed - 000001 000 011 010 000 001

VLC1 - 000100010001100010

VLC2 - 010010001

Total bits - $21 + 18 + 9$

$$= 48$$

of VLC2

Regroup the bits, in different manner

VLC2 - [0] [10] [0][1] [0] [0] [01] ... (Grouping)

AEABAAD - decoded message 1

Regroup single single bits

VL2 - [0] [1] [0] [0] [1] [0] [0] [0] [1]

ABAABAAAB - decoded message 2

• Prefix code:-

Prefix codes are the codes in which no code word forms the prefix of any other code

• Instantaneous code:-

As soon as sequence of bits corresponding to any one of the corresponding code word is detected, we can declare that symbol is decoded. This code is called instantaneous code

There is no decoding delay in instantaneous codes

- Q.1 Consider the following where source of size 4 has been encoded in binary codes with 0 and 1.
Identify different codes

n_i	code 1	code 2	code 3	code 4	code 5	code 6
n_1	00	00	0	0	0	1
n_2	01	01	1	10	01	01
n_3	00	10	00	110	011	001
n_4	11	11	11	111	0111	000

Types of code

- 1) Fixed code / length - code 1, code 2
- 2) Variable length - code 3, code 4, code 5, code 6
- 3) Instantaneous code - } 2, 4, 6
- 4) Prefix code -
- 5) Uniquely decodable -
- 6) Distinct Codes -

- Q.2 Following example has 2 binary code for 4 symbol each having probability 0.5, 0.25, 0.125, 0.125 resp Compare their efficiency

code length - No of bits for encoding. Avg code length is - . Average no of bits per source letter or symbol

$$= \bar{L} = \sum_{i=1}^N n_i P(n_i)$$

- code efficiency

$$\eta = \frac{H(s)}{L} - \text{entropy}$$

- γ = code redundancy

$$\gamma = 1 - \eta$$

code 1	n_i	code 2	n_i
00	2	0	1
01	2	10	2
10	2	110	3
11	2	111	3

n_i	$P(n_i)$
n_1	0.5
n_2	0.25
n_3	0.125
n_4	0.125

- Code 1

Avg length

$$L_{\text{code 1}} = n_1 p_1 + n_2 p_2 + n_3 p_3 + n_4 p_4$$

$$H(n)_{\text{code 1}} = \sum_{i=1}^4 P(n_i) \log_2 \left(\frac{1}{P(n_i)} \right) = 1.75 \text{ bits/symbol}$$

$$\eta = \frac{H(n)}{L}$$

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

$$\eta = \frac{1.75}{2} = 0.875$$

$$\eta \cdot 100\% = 87.5\%$$

- Optimum Code - If it is instantaneous and has minimum average length for given source for with the particular prob assignment for source symbol
- Entropy code :- When a variable length code is designed such that its average code per length L approaches the entropy of DMS (discrete memory less source)
- $H(n)$ for VLC2 (question continued)

$$L = n_1 p_1 + n_2 p_2 + n_3 p_3 + n_4 p_4 \\ 0.5 \times 1 + 0.25 \times 2 + 0.125 \times 3 \times 2$$

$$L = 1.75$$

$$H = 1.75$$

$$\therefore \eta = \frac{H(n)}{L}$$

$$\eta = \frac{1.75}{1.75}$$

$$\eta = 1, \eta \cdot 100\%.$$

- Q. A binary source is emitting an independent source of 0 and 1 with prob $P(0) = P, P(1) = 1 - P$. Plot the entropy of source vs P

$$\rightarrow H(n) = \sum_{i=1}^2 P_{ni} \log_2 \left(\frac{1}{P_{ni}} \right)$$

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

Ques Make graph

• Shannon-Fano coding

- 1) It is a source coding technique for constructing prefix code based on set of symbols and based on probabilities. This is sub-optimal code as it does not achieve lowest possible codeword length.
- 2) It produces fairly efficient variable length coding.
- 3) It is used in IMplode compression method which is part of zip file format.

• Steps of algorithm

- 1) Arrange the source or symbol in order of decreasing order. The symbols which equal probabilities can be listed in any arbitrary order.
- 2) Divide the set into 2, such that the sum of the probabilities in each set is same or nearly same.
- 3) Assign 0 to the upper set and 1 to the lower set.
- 4) Repeat steps 2 and 3 until each subset contains a single symbol.

Ques- A DMS has five symbols n_i , with their probabilities as follows. Construct the shanon-fano code for X. Calculate efficiency of code.

n_1	P_{n_1}
n_1	0.4
n_2	0.17
n_2	0.18
n_3	
n_4	0.1
n_5	0.15

→ Arranging

n_i	P_{n_i}	column 1	
n_1	0.4	0	0
n_3	0.18	0	1
n_2	0.17	1	0
n_5	0.15	1	1
n_4	0.1	1	1

Take the group of 2 such that sum may not cross column 2

column 3	code	codeword
0	00	2
1	01	2
0	10	2
0	110	3
1	111	3

$$L = P \times n$$

$$= 0.4 \times 2 + 0.18 \times 2 + 0.17 \times 2 + 0.15 \times 3 + 0.1 \times 3 \\ 0.8 + 0.36 + 0.34 + 0.45 + 0.3 = 2.25 \text{ bits/symbol}$$

$$H(X) = 0.4 \log_2 \left(\frac{1}{0.4} \right) + 0.18 \log_2 \left(\frac{1}{0.18} \right) + 0.17 \log_2 \left(\frac{1}{0.17} \right) + 0.15 \log_2 \left(\frac{1}{0.15} \right) \\ + 0.3 \log_2 \left(\frac{1}{0.3} \right) = 2.149 \text{ bits/symbol}$$

$$\eta = \frac{2.14}{2.25} = 0.955$$

$$\therefore \eta \% = 95.5\%$$

another way

n_i	p_{ni}	column 1	column 2	column 3	code	Length
n_1	0.4	0	0	0	00	2
n_2	0.18	1	0	0	100	3
n_3	0.17	1	0	1	101	3
n_4	0.15	1	1	0	110	3
n_5	0.1	1	1	1	111	3

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

$$H(n) = 2.149 \text{ bits/symbol}$$

$$\eta = \frac{2.14}{2.2} = 0.9727$$

$$\therefore \eta \% = 0.9727 \times 100$$

$$= 97.27\% - \text{code efficiency}$$

- Huffman coding

Huffman coding produces prefix code that always achieves the lowest possible avg code word length. It is an optimal code which has highest efficiency or lowest redundancy. Also known as optimal code and redundancy code.

- Procedure for Huffman coding :-

- 1) List the source code in decreasing prob. The symbols with equal prob can be arranged in any arbitrary order.
- 2) Combine prob. of symbol having the smallest prob. Now reorder the resultant prob. This process is called reduction 1. The same process is repeated until there are 2 ordered prob. remaining. The final step is called as last reduction.
- 3) Start encoding with the last reduction. Assign 0 as the first digit in the code words for all the source symbols, associated with the first prob of the last reduction. Then assign 1 to the second probability.
- 4) Now go back to the previous reduction step. Assign 0 and 1 to the ^{2nd} digit for the 2 prob that was combined in this reduction step. retaining all assignments made in step 3.

5) Repeat step 4 until first column is reached

6) Construct Huffman code for the following

n_1	P_{n_1}	
n_4	0.4	$\rightarrow 0.4$
n_2	0.18	
n_3	0.17	
n_4	0.15	
n_5	0.10	

combine these 2, 0.25,
Now arrange in decreasing
order again

Now take 0.18 and 0.17.

combined = 0.35

∴ Now decreasing order - 0.4

0.35

0.25

again do the same process

new combined = 0.6 - reduction 2

∴ Now decreasing order = 0.6

0.4

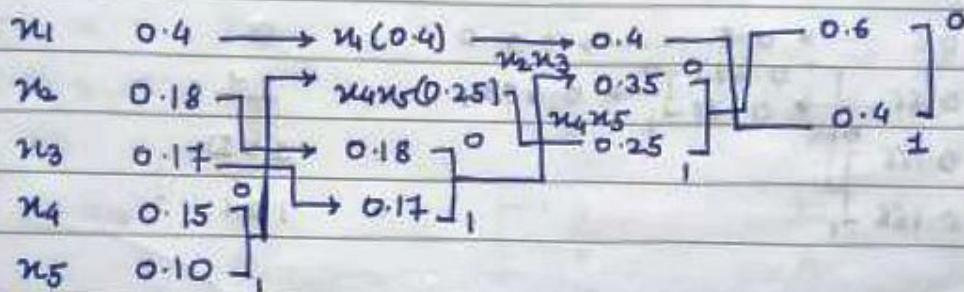
At last 2 probabilities will come - One highest and one lowest

Kraft's inequality condition

Kraft's inequality is a necessary and sufficient condition to prove existence of prefix code (uniquely decodable code) given n symbol and l_i are no of bits used to represent a symbol for all $i = 1$ to n

$$L = \sum_{i=1}^n 2^{-l_i} \leq 1$$

Corrected Method (Huffman code)



$n_1 - 1$

$n_2 - 000$

$n_3 - 001$

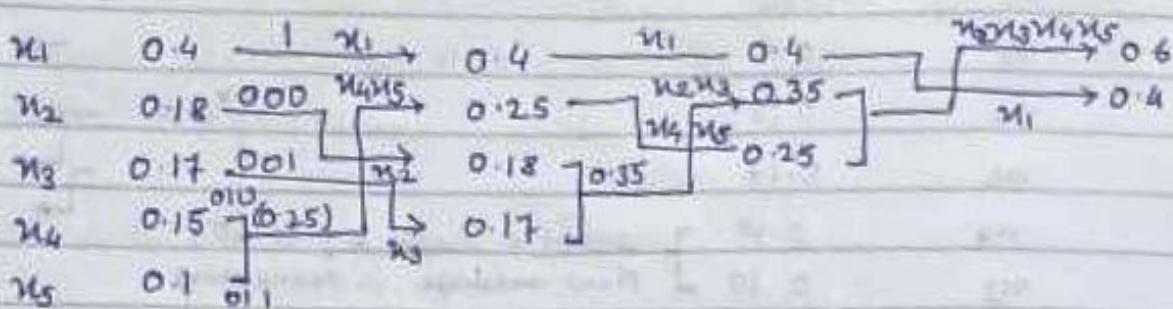
$n_4 - 010$

$n_5 - 011$

Construct Huffman code for the following symbols given

Huffman coding

$n_i \quad P_{n_i}$



n_i	P_{n_i}	code	code word length
n_1	0.4	1	1
n_2	0.18	000	3
n_3	0.17	001	3
n_4	0.15	010	3
n_5	0.1	011	3

$$H(n) = \sum_{i=1}^5 P_{n_i} \log_2 \left(\frac{1}{P_{n_i}} \right)$$

$$0.4 \log_2 \left(\frac{1}{0.4} \right) + 0.18 \log_2 \left(\frac{1}{0.18} \right) + 0.17 \log_2 \left(\frac{1}{0.17} \right)$$

$$+ 0.15 \log_2 \left(\frac{1}{0.15} \right) + 0.11 \log_2 \left(\frac{1}{0.11} \right)$$

$$H(n) = 2.151 \text{ bits/symbol}$$

$$\overline{L}_{\text{avg}} = \sum_{i=1}^5 n_i P_{n_i} = 2.2$$

$$\eta = \frac{2.151}{2.2} = 0.977 = 97.7\%$$

∴ The efficiency is 97.7%.

- "SAHARA" - mode with Huffman code

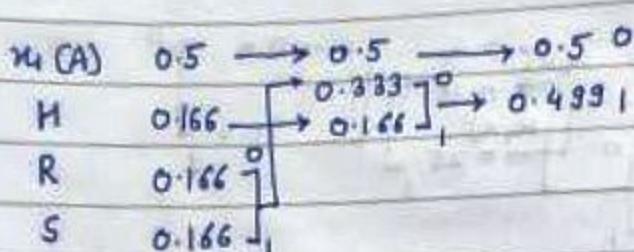
$$P(S) = \frac{1}{6}$$

$$P(A) = \frac{1}{2}$$

$$P(H) = \frac{1}{6}$$

$$P(R) = \frac{1}{6}$$

$$\pi_i \quad p_{ni}$$



	code	Code word
S	0	1
H	11	2
R	100	3
A	101	3

$$\begin{aligned}
 L &= 0.5 \times 1 + 0.166 \times 2 + 0.166 \times 6 \\
 &= 0.996 + 0.5 + 0.332 \\
 &= 1.828 \text{ bits/symbol}
 \end{aligned}$$

$$H = 1.055 \text{ bits/symbol}$$

$$\eta = \frac{1.055}{1.828}$$

$$\begin{aligned}
 &= 0.5771 \\
 &\approx 57.71\%
 \end{aligned}$$

∴ The efficiency is 57.71%.

Corrected solution

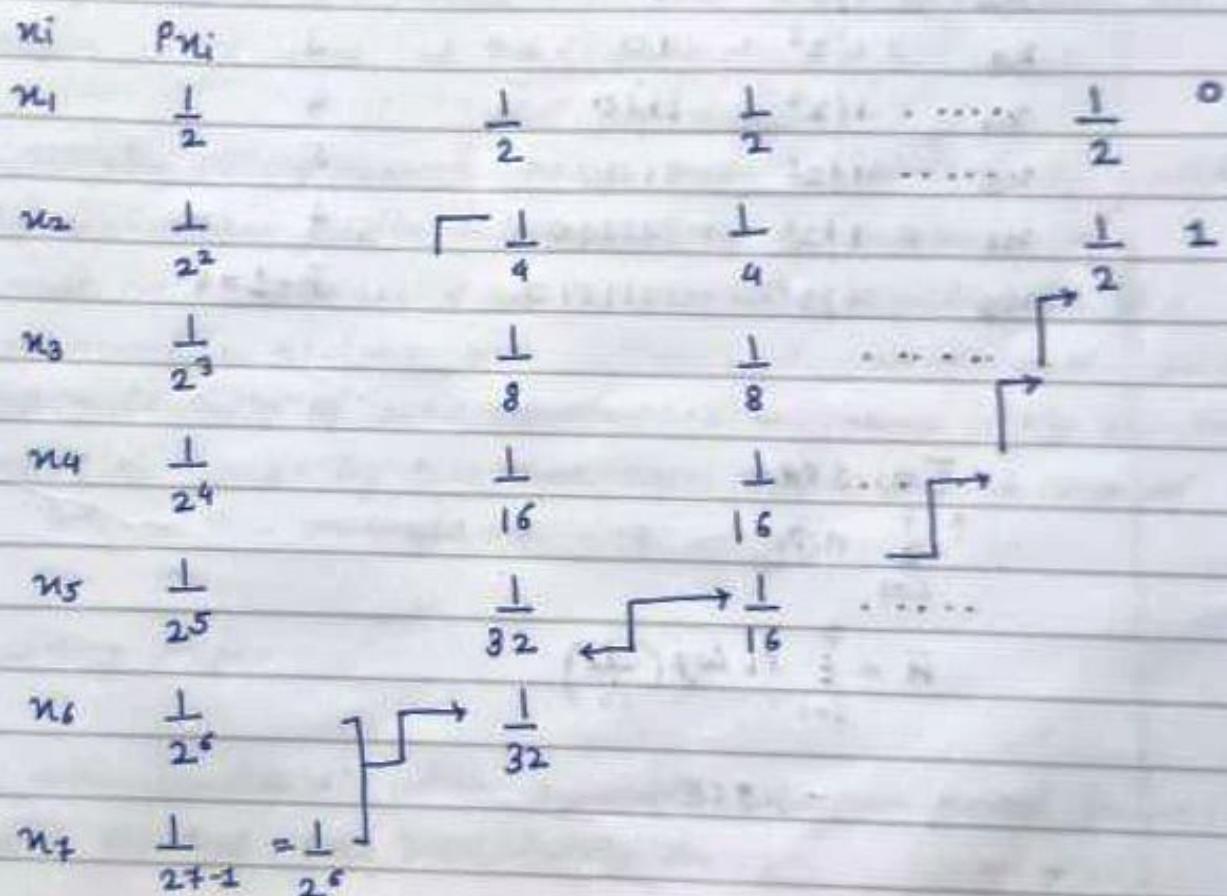
$$H = 1.73 \text{ bits/symbol}$$

$$L = 1.83 \text{ bits/symbol}$$

$$\eta = \frac{H}{L} = \frac{1.73}{1.83}$$

$$\eta = 97.8\%$$

Ex-A discrete memory less source has 7 signals



Construct huffman code and calculate efficiency of code
and comment about Kraft's inequality condition

- More self information of n_1 is 1, n_2 is 2, ...
 n_6 is 6
 : The information (self) in 2 bits transfer will be 1
 $H(\text{codeword length}) = \text{information transfer}$

KIT tells us about the following

n_i	p_{n_i}	code	codeword length
n_1	$1/2$	0	1
n_2	$1/2^2$	10	2
n_3	$1/2^3$	110	3
n_4	$1/2^4$	1110	4
n_5	$1/2^5$	11110	5
n_6	$1/2^6$	111110	6
n_7	$1/2^7$	1111110	$7 - 1 = 6$

$$\bar{L} = 1.968$$

$$\uparrow \sum_{i=1}^7 n_i p_i$$

$$\bar{H} = \sum_{i=1}^7 p_i \log \left(\frac{1}{p_i} \right)$$

$$= 1.968$$

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

$$\eta = \frac{1.968}{1.968}$$

$$\eta = 1$$

$$\eta \% = 100\%$$

∴ The efficiency is 100%.

- Key point

- Huffman codes are only optimal if the probabilities of symbol are -ve power of 2. This is because all the prefix codes were at last level
- Arithmetic coding doesn't have this restriction. It works by representing file to be encoded by an interval of real no.'s between 0 and 1; successive symbol in the message reduce this interval or advanced width of the probability of that symbol. The more likely symbols reduce range by less and thus add fewer no. of bits to the message.
- Encoding steps:

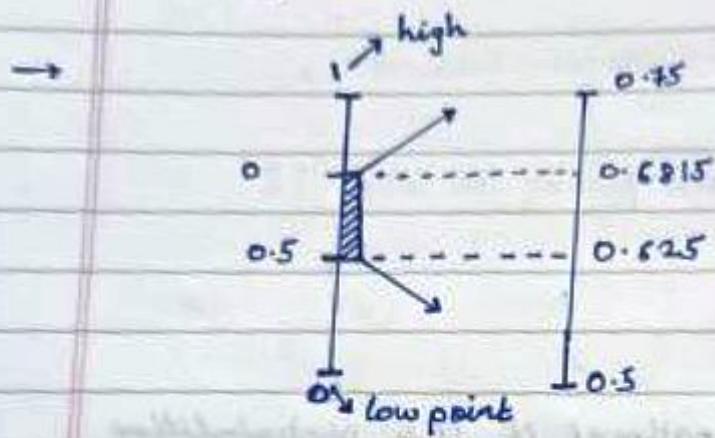
- i) To code symbol 's' when symbols are numbered from 1 to n and symbol s has probability $Pr[i]$.

$$\text{low-bound} = \sum_{i=1}^{s-1} Pr[i] \quad \text{range} = \text{high-low}$$

$$(lao) = \text{low} \quad (\text{range} + \text{low-bound})$$

$$\text{high-bound} = \sum_{i=1}^s Pr[i], \quad \text{high} = \text{low} \quad (\text{range} + \text{high-bound})$$

Eg:- Consider symbols 'A', 'B', 'C' with probabilities 0.5, 0.25 and 0.25. Encode given input symbol stream BAC using Arithmetic code



For B

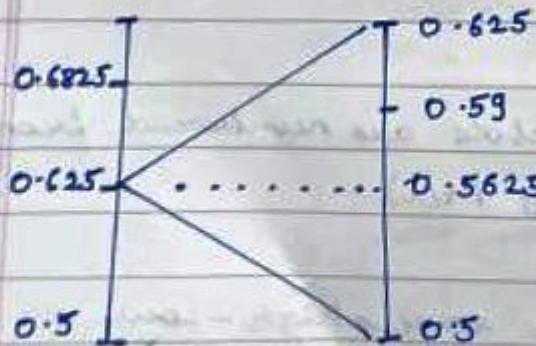
$$0.5 + (0.75 - 0.5) \times 0.5 = 0.625 \dots \text{1st step}$$

↓ ↓ ↓ ↓
 low high low Prob.

$$= 0.625 + (0.75 - 0.625) \times 0.25 = 0.6875 \dots \text{2nd step}$$

$$\cdot \quad 0.6875 + (0.75 - 0.6875) \times 0.25 = 0.75 \dots \text{3rd step}$$

Here we take with each alphabet's probability are by one and the output / answer of first step is the first value of the second step



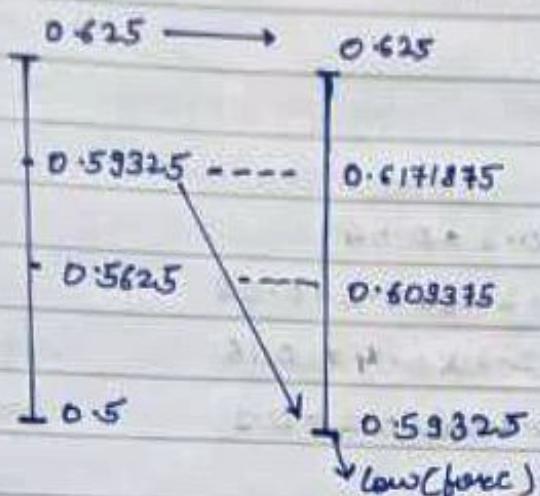
For A

$$0.5 + (0.625 - 0.5) \times 0.5 = 0.5625$$

$$0.5625 + (0.625 - 0.5) \times 0.25 = 0.59375$$

$$0.59375 + (0.625 - 0.5) \times 0.25 = 0.625$$

Now after doing for first letter observe the answer / output reading and according to make diagram and decide high and low values for next letter



For C

$$0.59375 + (0.625 - 0.59375) \times 0.5$$

$$= 0.609375$$

$$0.609375 + (0.625 - 0.609375) \times 0.25$$

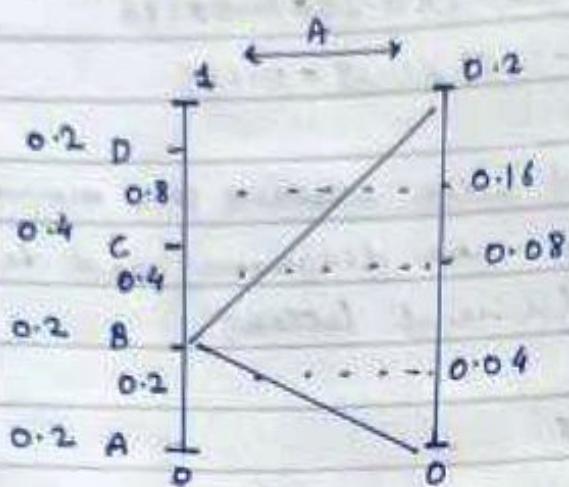
$$= 0.6171875$$

$$0.6171875 + (0.625 - 0.6171875) \times 0.25,$$

$$= 0.625$$

The final answer lower value of last letter
∴ Final Answer = "0.59375"

- Q. Four symbols A, B, C, D with probabilities 0.2, 0.2, 0.4, 0.2. Transmitted message is ABC



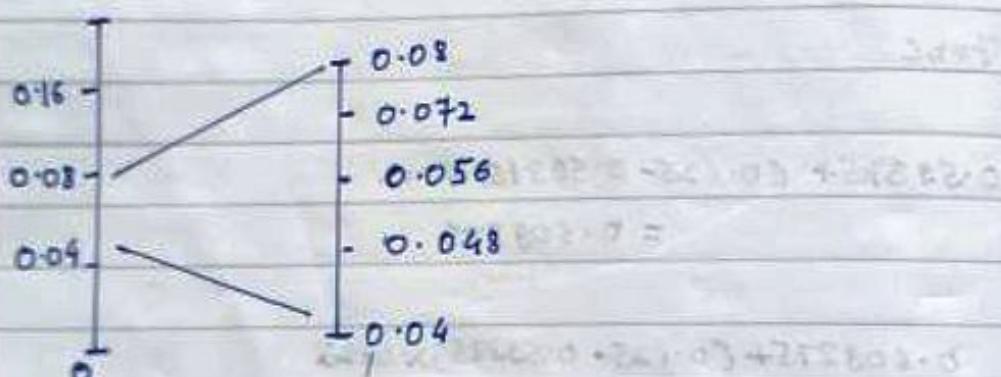
For A

$$0 + (0.2 - 0) \times 0.2 = 0.04$$

$$0.04 + (0.2 - 0) \times 0.2 = 0.08$$

$$0.08 + (0.2 - 0) \times 0.4 = 0.16$$

$$0.16 + (0.2 - 0) \times 0.2 = 0.2$$

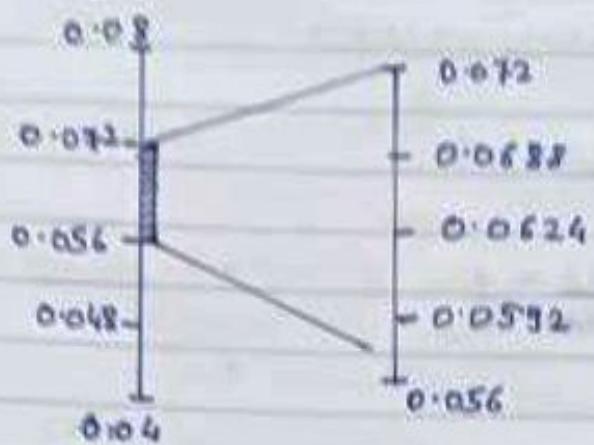


For B

$$0.04 + (0.08 - 0.04) \times 0.2 = 0.048$$

$$0.048 + (0.08 - 0.04) \times 0.2 = 0.056$$

$$0.056 + (0.08 - 0.04) \times 0.4 = 0.072$$



For C

$$0.056 + (0.072 - 0.056) \times 0.2 = 0.0592$$

$$0.0592 + (0.072 - 0.056) \times 0.2 = 0.0624$$

$$0.0624 + (0.072 - 0.056) \times 0.4 = 0.0688$$

The final Answer is low value of last letter

Final Answer = 0.056

Q COFFEE

$$C = 1/6, O = 1/6, F = 2/6, E = 2/6$$

Solve this by the above method. We can take letters in any order

Runlength coding

Random bits - 1111111110000000111

make groups

12 (1's), 7 (0's), 3 (1's)

represent no in bits

$(1100, 1), (111, 0), (11, 1) \rightarrow 12\text{-bits}$

original bits = 22 bits

$$\text{Compression} = \frac{12}{22} = \frac{6}{11}$$

- Lempel-Ziv-Walsh (LZW coding)
- Dictionary based coding algorithm

Ex: Encode using LZW - "DRESSES"

original dictionary

symbol	code
D	0
R	1
E	2
S	3

input	current string	seen before	encoded o/p	New dict entry/o/p
D	D	Yes	-	-
D R	DR	No	0	DR 4
DR E	RE	No	01	RE 5
DRE S	ES	No	01, 2	ES 6
DRES S	SS	No	0, 1, 2, 3	SS 7
DRESS E	SE	No	0, 1, 2, 3, 3	SE 8
DRESSE S	ES	Yes	-	-
ES	ES	Yes	-	-
		Yes	0, 1, 2, 3, 3, 6	-
Final code - 0, 1, 2, 3, 3, 6				

Corrected answer (DADDA - DAADAA)

original dictionary

symbol	code
D	0
A	1
-	2

Input	current string	seen before	encoded o/p	New dictionary entry/index
D	D	Yes	-	-
D A	DA	No	0	DA/3
DA/D	AD	No	0,1	AD/4
DAD A	DA	Yes	-	-
DADA -	DA-	No	0,1,3	DA- /5
DADA-D	-D	No	0,1,3,2	-D /6
DADA-D/A	DA	Yes	-	-
DADA-DA/A	DAA	No	0,1,3,2,3	DAA/7
DADA-DAAD/D	AD	Yes	-	-
DADA-DAAD/D	ADD	No	0,1,3,2,3,4	ADD/8
DADA-DAADD/A	DA	Yes	-	-
-	DA	-	0,1,3,2,3,4,3	-

Final Answer - 0,1,3,2,3,4,3

Flow chart



Ex-1. Throw a die and obs no of dots appearing on the front face

i) construct sample space

$$\rightarrow S = \{1, 2, 3, 4, 5, 6\}$$

ii) let A be an event that odd no. occurs

$$\rightarrow \{1, 3, 5\} = S(A)$$

iii) let B be an event that even no. occurs

$$\rightarrow S = \{2, 4, 6\} (B)$$

iv) let C be an event that prime no. occurs

$$\rightarrow S = \{2, 3, 5\}$$

v) find the event that an even or prime no. occurs

$$\rightarrow S = \{2, 3, 4, 5, 6\}$$

vi) list the outcomes for an event that even prime no. occurs

$$\rightarrow B \cap C = \{2\}$$

vii) find the event that prime does not occur

$$\rightarrow C' = \{1, 4, 6\}$$

viii) find the event that seven dots appear on top face

$$\rightarrow S = \{\emptyset\}$$

ix) find the event that even and odd no. occurs simultaneously

$$\rightarrow A \cap B = \emptyset$$

Ex-2-

	employed	Unemployed	Total
Male	250	50	300
Female	150	100	250
Total	400	150	550

a) If a person is male, what is the prob that he is unemployed

- $0 \leq P(A) \leq 1$
- $P(\bar{A}) = 1 - P(A)$
- $P(A \cup B) = P(A) + P(B)$
- $P(\emptyset) = 0$
- $P(A \cup B) = P(A) + P(B) - P(A \cap B)$
- $P(B|A) = P(A \cap B) / P(A)$

$$\rightarrow a) \frac{50}{300} = \frac{1}{6} \rightarrow \text{probability} \quad \therefore \left[P(U|M) = \frac{P(U \cap M)}{P(M)} = \frac{50}{300} = \frac{1}{6} \right]$$

- Conditional Probability - "cond" prob of an event A given that B has happened, it is defined as

$$P(A|B) = \frac{P(A \cap B)}{P(B)} ; P(B) > 0$$

$\therefore P(A \cap B) \rightarrow$ joint probability of A and B

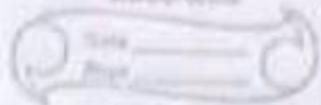
$$P(B|A) = \frac{P(A \cap B)}{P(A)} ; P(A) > 0$$

$$P(A \cap B) = P(A|B) \cdot P(B) = P(B|A) \cdot P(A)$$

b) If a person is female, what is prob that she is employed

$$P(E|F) = \frac{P(E \cap F)}{P(F)}$$

$$= \frac{150}{250}$$



- c) If a person is employed, what is probability he is male

$$P(M|E) = \frac{P(M \cap E)}{P(E)}$$

- Independent event :-

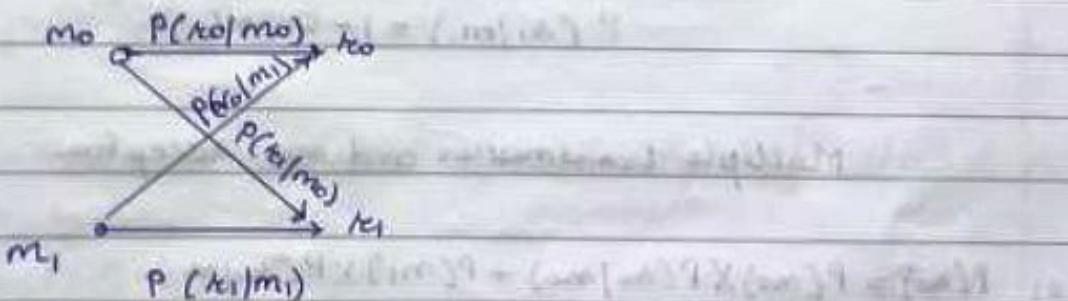
Two events A and B are statistically independent if $P(A|B) = P(A)$ and $P(B|A) = P(B)$

i.e. occurrence or non-occurrence b or A are said to be independent

For independent event $P(A \cap B) = P(A) \cdot P(B)$.

[as $P(A|B) = P(A)$ for independent event]

- In a binary communication system either 0 or 1 is transmitted. Due to channel noise, a one can be received as zero and vice-versa. Let m_0 and m_1 denote event of transmitting 0 and 1 respectively. Let r_0 and r_1 denote event of receiving 0 and 1 respectively.



Given $P(m_0) = 0.5$

$P(r_1|m_0) = 0.2$

$P(r_0|m_1) = 0.1$

Find

a) $P(r_0), P(r_1)$

b) If a 1 was received, what is probability that a 1 was sent?

c) If a 0 was received, what is probability that 0 was sent

$$P(m_0|r_0) = ?$$

d) Calculate probability of error

e) Calculate probability of error of Pe that transmitted signal correctly reaches to receiver

$$\rightarrow \text{Sol}^n: \quad P(m_0) = 0.5$$

$$P(r_1|m_0) = 0.2$$

$$P(r_0|m_1) = 0.1$$

$$P(m_1) = 1 - P(m_0) = 0.5$$

$$P(r_0|m_0) + P(r_1|m_0) = 1$$

$$\therefore P(r_0|m_0) = 1 - P(r_1|m_0)$$

$$1 - 0.2 = 0.8$$

$$P(r_0|m_1) + P(r_1|m_1) = 1$$

$$P(r_1|m_1) = 1 - 0.1 = 0.9$$

Multiple transmission and one reception

a) $P(r_0) = P(m_0) \times P(r_0|m_0) + P(m_1) \times P(r_0|m_1)$

$$0.5 \times 0.8 + 0.5 \times 0.1$$

$$= 0.45$$

$$\begin{aligned}
 P(A_1) &= P(V_1|m_0) \times P(m_0) + P(V_1|m_1) \times P(m_1) \\
 &= 0.2 \times 0.5 + 0.9 \times 0.5 \\
 &= 0.55
 \end{aligned}$$

$$\begin{aligned}
 b) P(m_1|A_1) &= \frac{P(A_1|m_1) \cdot P(m_1)}{P(A_1)} \\
 &= \frac{0.9 \times 0.5}{0.55} \\
 &= 0.818
 \end{aligned}$$

$$\begin{aligned}
 c) P(m_0|A_0) &= \frac{P(A_0|m_0) \times P(m_0)}{P(A_0)} \\
 &= \frac{0.8 \times 0.5}{0.45} \\
 &= 0.889
 \end{aligned}$$

Total probability

Let A_1, A_2 be usually exclusive. $[A_i \cap A_j = \emptyset]$. Now B is any event in S

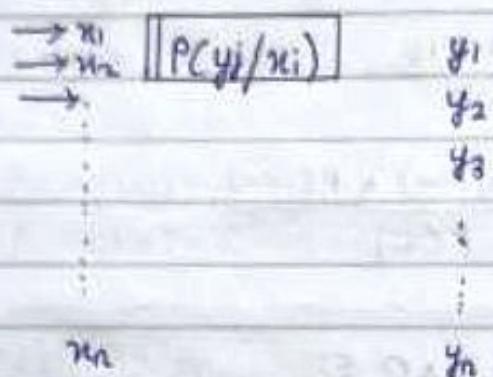
$$\begin{aligned}
 \text{Then } P(B) &= \sum_{i=1}^n P(B \cap A_i) \\
 &= \sum_{i=1}^n P(B|A_i) P(A_i)
 \end{aligned}$$

$$P(A_i|B) = \frac{P(B|A_i) P(A_i)}{\sum_{i=1}^n P(B|A_i) P(A_i)}$$

Channel Model

A communication channel is a transmission path through which symbols flow through the receiver.

The -- channel is discrete because alphabets X and Y both are finite. Hence channel is discrete. Since present off of channel solely depends on the current I/P and not on input of previous stages.



$X = \text{input}$

$Y = \text{output}$

$P(Y_j/X_i) \rightarrow \text{channel transition probability}$

case 1

$$[P(Y|X)] = \begin{bmatrix} P(Y_1/x_1) & P(Y_2/x_1) & P(Y_3/x_1) & \dots & P(Y_n/x_1) \\ P(Y_1/x_2) & P(Y_2/x_2) & P(Y_3/x_2) & \dots & P(Y_n/x_2) \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ P(Y_1/x_m) & P(Y_2/x_m) & P(Y_3/x_m) & \dots & P(Y_n/x_m) \end{bmatrix}$$

i) Each row contains conditional prob of all off symbols.

$$\sum_{j=1}^n P(y_j|x_i) = 1 \text{ for all } i$$

ii) The input probabilities $P(n)$ are given by a row matrix

$$3) [P(Y)] = [P(Y_1) \ P(Y_2) \ \dots \ P(Y_n)]$$

$$[P(Y)] = [P(x)]^T [P(Y|x)]$$

If $[P(x)]$ is represented as diagonal matrix

$$[P(x)_i] = \begin{bmatrix} P(x_1) & 0 & 0 & \dots & 0 \\ 0 & P(x_2) & 0 & \dots & 0 \\ 0 & 0 & P(x_3) & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & P(x_m) \end{bmatrix}$$

$$\text{Then } [P(x, Y)] = [P(n)]_d \cdot [P(Y|n)]$$

$[P(x, Y)]$ matrix is of form $P(n_i, y_j)$

↓

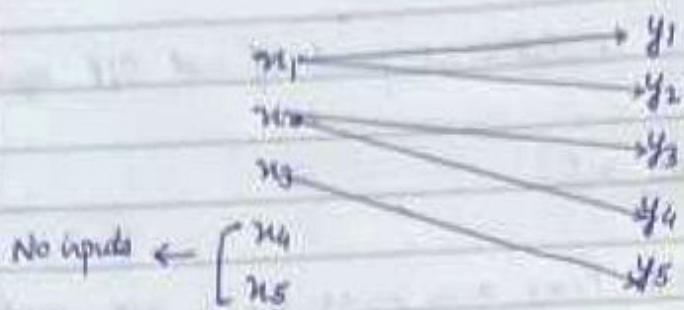
It is called joint probability matrix
and $P(n_i, y_j)$ is called joint probability of sending
 n_i and receiving y_j .

• Special channel

A) A lossless channel

$$(y_1|x_1) \ (y_2|x_1) \ (y_3|x_1) \ (y_4|x_1) \ (y_5|x_1)$$

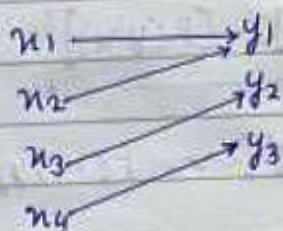
$$[P(Y|x)] = \begin{bmatrix} 1/6 & 5/6 & 0 & 0 & 0 \\ (y_1|x_2) & 0 & 1/4 & 3/4 & 0 \\ (y_1|x_3) & 0 & 0 & 0 & 1 \end{bmatrix}$$



If channel matrix of a channel contains only one non-zero element in the column, then that channel is defined as lossless channel.

B) Deterministic Channel

$$[P(Y|X)] = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



for deterministic channel, the channel matrix contains only 1 non zero elements in each row.

Since each row has one non-zero element this element must be unity.

When a particular input symbol is transmitted through this channel it is known which output symbol is received

c) Noiseless channel

If a channel is noisy, if it is both deterministic and lossy

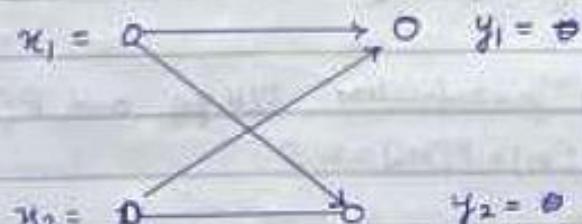
$$P(Y|X) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$



d) Binary symmetric channel (BSC)

$$[P(Y|X)] = \begin{bmatrix} 1-p & p \\ p & 1-p \end{bmatrix}$$

It has two inputs and two outputs : Channel is symmetric as probability of misinterpreting a transmitted as D as 1 is same as misinterpreting 1 as 0



$$x_1 = 0, y_1 = 0$$

$$x_1 = 1, y_1 = 1$$

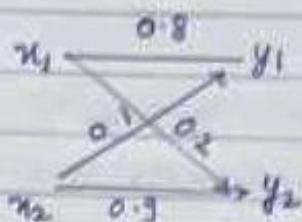
$$x_2 = 0, y_2 = 1$$

$$x_2 = 1, y_2 = 0$$

$$P(u_1) = P(u_2) = 0.5$$

a. For the binary channel shown, a) find channel matrix

$$b) P(y_1), P(y_2) \text{ when } P(u_1) = P(u_2) = 0.5$$



$$\rightarrow P[Y|X] = \begin{bmatrix} 0.8 & 0.2 \\ 0.1 & 0.9 \end{bmatrix}$$

$$[P(X)] = [0.5 \quad 0.5]$$

$$[P(X)] = [0.5 \quad 0.5] [P(Y|X)]$$

$$[P(X)] = [0.5 \quad 0.5] \begin{bmatrix} 0.8 & 0.2 \\ 0.1 & 0.9 \end{bmatrix}$$

$$[P(Y)] = [0.45 \quad 0.55]$$

Find the joint probabilities $P(u_1, y_2)$ and $P(u_2, y_1)$
when $P(u_1) = P(u_2) = 0.5$

$$[P(X, Y)] = [P(X)] \cdot [P(Y|X)]$$

$$\begin{bmatrix} P(u_1, y_1) & P(u_1, y_2) \\ P(u_2, y_1) & P(u_2, y_2) \end{bmatrix} = \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} 0.8 & 0.2 \\ 0.1 & 0.9 \end{bmatrix} = \begin{bmatrix} 0.40 & 0.10 \\ 0.05 & 0.45 \end{bmatrix}$$

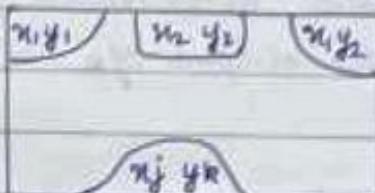
Joint entropy and conditional entropy



$$S_1 = n_1, n_2, \dots, n_m$$



$$S_2 = y_1, y_2, \dots, y_n$$



→ Rough representation

$$P(X) = P(n_j)$$

$$P(Y) = P(y_k)$$

$$P(X, Y) = [P(n_j, y_k)]$$

$$\therefore P(B) = \sum_{i=1}^n P(B \cap A_i) = \sum_{i=1}^n P(B|A_i)P(A_i)$$

m

$$H(n) = -\sum_{j=1}^m P(n_j) \log_2 P(n_j)$$

$$H(y) = -\sum_{k=1}^n P(y_k) \log_2 P(y_k)$$

$$H(n, y) = -\sum_{j=1}^m \sum_{k=1}^n P(n_j, y_k) \log_2 P(n_j, y_k)$$

H(n) and H(y) are called marginal entropy of n and y respectively. The entropy of input symbols n, before their transmission is called as "priori" entropy.

H(X, Y) is joint entropy of (X, Y)

Based on conditional probabilities we can calculate entropies

$$H(X, Y_k) = - \sum_{j=1}^n P(x_j | y_k) \log_2 P(x_j | y_k)$$

The entropy of the input symbol after transmission and reception of particular output y_k is called "posterioric" entropy

$$H(X, Y) = \sum_{k=1}^m P(y_k) H(X | y_k)$$

Equivocation:

It is defined as average value of all conditional probability in above equation where $k=1$ to n

$$\text{formula} = - \sum_{k=1}^n P(y_k) = \sum_{j=1}^m P(x_j | y_k) \log_2 P(x_j | y_k)$$

$$= - \sum_{j=1}^m \sum_{k=1}^n P(y_k) \cdot P(x_j | y_k) \log_2 P(x_j | y_k)$$

$$H(X|Y) = - \sum_{j=1}^m \sum_{k=1}^n P(x_j, y_k) \log_2 P(x_j | y_k) \quad \dots \textcircled{1}$$

$$H(Y|X) = - \sum_{j=1}^m \sum_{k=1}^n P(x_j, y_k) \underbrace{\log_2 P(y_k | x_j)}_{\text{change}} \quad \dots \textcircled{2}$$

Equivocation of $(X|Y)$

① It is the measure of $H(X|Y)$. It is the equivocation of n w.r.t y . It is the measure of uncertainty about the channel input. It describes how well one can recover the transmitted symbols from received symbols.

Equivocation of y w.r.t n

Avg uncertainty of channel o/p when n was transmitted and it describes how well one can recover the received symbols from transmitted symbols. It is measure of error or noise.

Mutual Information

$P(n_j) \rightarrow$ It is the prob that n_j would be selected for transmission. This is priori probability.

Uncertainty in this will be $-\log_2 P(n_j)$. Once y_k is received at the o/p, the state of knowledge about n_j is condⁿ probability, $P(n_j|y_k)$ which is called posterioric prob.

Uncertainty in this $\rightarrow -\log_2 P(n_j|y_k)$

- $I(n_j, y_k) = \text{initial uncertainty} - \text{final certainty}$
- Net reduction in uncertainty

$$I(n_j, y_k) = -\log_2 P(n_j) - [-\log_2 P(n_j|y_k)]$$

$$= -\log_2 P(n_j) + \log_2 P(n_j|y_k)$$

$$= \log_2 \left(\frac{P(n_j|y_k)}{P(n_j)} \right)$$

$$= \log_2 \left[\frac{P(n_j | y_k)}{P(n_j)} \right]$$

$$= \log_2 \left[\frac{P(n_j, y_k)}{P(n_j)P(y_k)} \right]$$

- Q. A binary channel has following properties $[P(Y|X)]$
- $$\begin{bmatrix} 1/3 & 2/3 \\ 2/3 & 1/3 \end{bmatrix}$$

$$P(n_1) = 1/4, P(n_2) = 3/4$$

Find $H(X), H(Y), H(X, Y), H(Y|X)$

$$[P(Y)] = [P(n)] \cdot [P(Y|X)]$$

$$[P(Y_1) \cdot P(Y_2)] = \left[\frac{1}{4} \quad \frac{3}{4} \right] \begin{bmatrix} 1/3 & 2/3 \\ 2/3 & 1/3 \end{bmatrix}$$

$$= \begin{bmatrix} 7/12 & 5/12 \end{bmatrix}$$

$$H(X) = -\sum_{j=1}^2 P(n_j) \log_2 P(n_j) = \frac{1}{4} \log_2 4 + \frac{3}{4} \log_2 \frac{4}{3}$$

$$= \frac{2}{4} + \frac{3}{4} \times 0.41 = 0.811 \text{ bits/msg}$$

$$H(Y) = -\sum_{k=1}^2 P(y_k) \log_2 P(y_k) = \frac{7}{12} \log \frac{12}{7} + \frac{5}{12} \log_2 \frac{12}{5}$$

$$= 0.976 \text{ bits/msg}$$

$$[P(X,Y)] = [P(X)] \otimes [P(Y|X)]$$

$$= \begin{bmatrix} 1/4 & 0 \\ 0 & 3/4 \end{bmatrix} \begin{bmatrix} 1/3 & 2/3 \\ 2/3 & 1/3 \end{bmatrix}$$

$$= \begin{bmatrix} 1/12 & 2/12 \\ 6/12 & 3/12 \end{bmatrix}$$

$$H(X,Y) = \sum P(X,Y) \log_2 P(X,Y)$$

$$\begin{aligned} &= \frac{1}{12} \log_2 12 + \frac{6}{12} \log_2 \frac{12}{6} + \frac{2}{12} \log_2 \left(\frac{12}{2}\right) \\ &\quad + \frac{3}{12} \log_2 \left(\frac{12}{3}\right) \\ &= 1.7299 \text{ bits/msg} \end{aligned}$$

Q In a column² system, a transmitter has 3 i/p symbols

$$A = \{a_1, a_2, a_3\}$$

$$B = \{b_1, b_2, b_3\}$$

The matrix below gives shows TPM with some probabilities

$$[P(A, B)] = \begin{bmatrix} 1/12 & * & 5/36 \\ 5/36 & 1/3 & 5/36 \\ * & 1/6 & * \end{bmatrix}$$

$$[P(B)] = [1/3 \quad 14/36 \quad *]$$

a) Find missing probabilities (*) in the table

b) Find $P(b_3|a_1)$, $P(a_2|b_3)$

c) Are the events a_1 and b_1 statistically independent? Why?

→ Property of JPM:-

i) By adding all elements of JPM column wise we can obtain probability of o/p symbols

$$\sum_{i=1}^q P(a_i, b_j) = P(b_j)$$

ii) By adding all elements of JPM row wise we can obtain probability of i/p symbols

$$\sum_{j=1}^q P(a_i, b_j) = P(a_i)$$

iii) The sum of all elements of JPM is equal to 1

$$\rightarrow a) P(b_1) + P(b_2) + P(b_3) = 1$$
$$\frac{1}{3} + \frac{14}{36} + n = 1$$
$$n = 1 - \frac{1}{3} - \frac{14}{36} = \frac{5}{18}$$

	b_1	b_2	b_3
a_1	$\frac{1}{12}$	$\frac{1}{9}$	$\frac{5}{36}$
a_2	$\frac{5}{36}$	$\frac{1}{9}$	$\frac{5}{36}$
a_3	$\frac{1}{9}$	$\frac{1}{6}$	0

$$[P(B)] \quad \frac{1}{12} \quad \frac{14}{36} \quad \frac{5}{18}$$

Using JPM property 1

$$P(a_1, b_1) + P(a_2, b_1) + P(a_3, b_1) = P(b_1)$$

$$\frac{1}{12} + \frac{5}{36} + n = \frac{1}{3}$$

$$n = \frac{1}{9}$$

$$P(a_1, b_2) + P(a_2, b_2) + P(a_3, b_2) = P(b_2)$$

$$\begin{aligned}P(a_1, b_2) &= \frac{14}{36} - \frac{1}{9} - \frac{1}{6} \\&= \frac{1}{9}\end{aligned}$$

$$P(a_3, b_3) = 0$$

$$P(a_1, b_1) + P(a_2, b_2) + P(a_3, b_3) = P(a_1)$$

$$P(a_1) = \frac{1}{3}$$

$$P(a_2) = \frac{7}{18}$$

$$P(a_3) = \frac{5}{18}$$

b) $P(b_3 | a_1) = \frac{P(a_1, b_3)}{P(a_1)} = \frac{5}{12}$

$$P(a_1 | b_3) = \frac{P(a_1, b_3)}{P(b_3)} = \frac{5}{8}$$

c) $P(a_1) \cdot P(b_1) = P(a_1, b_1)$

$$\frac{1}{3} \times \frac{1}{3} \neq \frac{1}{12}$$

It is not statistically independent

- V. Imp :-

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

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

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

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

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

- For lossless channel :-
channel capacity $c = \log_2 M$ ↗ if p symbol

- For deterministic
 $c = \log_2 n$ ↗ q symbols

- For Noiseless:-

$$c = \log_2 m = \log_2 n$$

- For BSC

$$c = 1 + p \log_2 p + (1-p) \log_2 (1-p)$$