

: Information theory:

Information source may be an object which produces an event, the outcome of which is selected at random according to probability distribution. The analog source can be transformed to discrete source through the use of sampling & quantization techniques. The set of source involves is called the source alphabet and the elements of this set are called symbols or letters.

Information content of a discrete memory less source properties

- 1) Information should be proportional to uncertainty of outcome
- 2) Information containing independent outcomes should add to get overall information code.

Information content of symbol [logarithmic measure of information] Let us consider a discrete memory less source (PMS) $X = \{x_1, x_2, x_3, \dots, x_m\}$. The information content of a symbol x_i is denoted by $I(x_i)$ is defined by $I(x_i) = \log_b \frac{1}{P(x_i)}$

$$\Rightarrow I(x_i) = -\log_b P(x_i)$$

$P(x_i) \rightarrow$ probability of occurrence of x_i

The unit $I(x_i)$ is the binary unit or $b=2$

$$I(x_i) = \frac{-\log p(x_i)}{\log 2}$$

(Q) A source produces one of 4 possible symbols during each interval having probabilities. $p(x_1) = \frac{1}{2}$, $p(x_2) = \frac{1}{4}$, $p(x_3) = \frac{1}{8}$, $p(x_4) = \frac{1}{8}$. Obtain the information content of each symbol?

$$\rightarrow I(x_1) = -\log_2 p(x_1)$$

$$= -\frac{\log(1/2)}{\log 2} = \underline{1 \text{ bit}}$$

$$I(x_2) = -\log_2 p(x_2)$$

$$= -\frac{\log(1/4)}{\log 2} = \underline{2 \text{ bits}}$$

$$I(x_3) = -\log_2 p(x_3)$$

$$= -\frac{\log(1/8)}{\log 2} = \underline{3 \text{ bits}} = I(x_4)$$

This reveals the fact that if probability of occurrence is less this information carried is more & vice versa.

(Q) In a binary PCM if zero occurs with probability $1/4$ & '1' occur probability $3/4$ then calculate the amount of info carried by each beneath?

$$\rightarrow I(x_0) = -\log_2 p(x_0) \quad I(x_1) = -\log_2 p(x_1)$$

$$= -\frac{\log(1/4)}{\log 2} = \underline{2 \text{ bits}} \quad I(x_1) = -\frac{\log(3/4)}{\log 2}$$

$$= 0.935 \text{ bit}$$

Entropy (Average information):

In practical communication systems we usually transmit long sequences of symbols from a information source thus we are more interested in Avg. info. that a source produces than the info. content of a single symbol, ~~in order to get the info. content of symbol~~. The avg. value of $I(x_i)$ over the alphabet of source x of m different symbols is given by

$$H(x) = E[I(x_i)] = \sum_{i=1}^m p(x_i) I(x_i)$$

↓
Entropy

$$H(x) = - \sum_{i=1}^m p(x_i) \log_2 p(x_i) \text{ bits/symbol}$$

The quantity $H(x)$ is called the entropy of source x if it is the measure of the avg info. content per source symbol.

Information Rate:

The rate at which source 'x' emits symbols is ' r ' (symbol/sec)

\Rightarrow the information Rate (R) of the source (x) is given by

$$R = r H(x) \text{ b/s} \quad r \rightarrow \text{rate at which symbol are generated}$$

$H(x) \rightarrow \text{Avg info. / Entropy (b/symb)}$

- (a) Discrete memoryless Source (DMS) X has 4 symbols x_1, x_2, x_3, x_4 with probabilities $0.4, 0.3, 0.2, 0.1$ resp. calculate $H(x)$. Find the amount of information contained in the message $x_1, x_2, x_1, x_3 \in X^4$ compare with the $H(x)$ obtained in part (i)

$$\begin{aligned} \rightarrow H(x) &= - \sum_{i=1}^m p(x_i) \log_2 p(x_i) \\ &= - \left[0.4 \cdot \frac{\log 0.4}{\log 2} + 0.3 \cdot \frac{\log 0.3}{\log 2} + 0.2 \cdot \frac{\log 0.2}{\log 2} + 0.1 \cdot \frac{\log 0.1}{\log 2} \right] \\ &= - (-0.528, -0.521, -0.464, -0.332) \\ &= \underline{1.845 \text{ bits/symbol}} \end{aligned}$$

$$\begin{aligned} I(x_1 x_2 x_1 x_3) &= -\log_2 P(x_1 x_2 x_1 x_3) \\ &= -\log_2 P(0.0048) = -\frac{\log(0.0048)}{\log 2} \\ &= \underline{7.702 \text{ bits}} \end{aligned}$$

$$I(x_1 x_3 x_3 x_2) = -\log_2 P(0.0012) = \underline{9.702 \text{ bits}}$$

- (b) Discrete source emits 5 symbols once in every millisecond. with probabilities $1/2, 1/4, 1/8, 1/16, 1/16$ resp. determine the source entropy and information rate?

$$\begin{aligned} \rightarrow H(x) &= - \sum_{i=1}^{\infty} p(x_i) \log_2 p(x_i) \\ &= - \left(\frac{1}{2} \frac{\log(1/2)}{\log 2} + \frac{1}{4} \frac{\log(1/4)}{\log 2} + \frac{1}{8} \frac{\log(1/8)}{\log 2} \right. \\ &\quad \left. + \frac{1}{16} \frac{\log(1/16)}{\log 2} \right) \end{aligned}$$

$$= -(-0.5 + (-0.5) - \frac{3}{8} - 0.5)$$

$$= \underline{1.475 \text{ bits/symbol}}$$

$$P(X) = P(x_1 x_2 x_3 \dots x_{15}) \\ = P(8.103 \times 10^{-2})$$

$$I(X) = 2 \log P$$

$$R = \gamma H(x) = 1.875 \text{ kbit/s}$$

$$P = \frac{1}{10^{-3}} H(x)$$

- Q) The probability of five possible experiments are given as
 $P(x_1) = 0.5, P(x_2) = 1/4, P(x_3) = 1/8, P(x_4) = P(x_5) = 1/16$. Determine the entropy and Info rate if there are 16 outcomes /sec.
 $\rightarrow H(x) = 1.875$

$$R = \gamma H(x) \\ = 16 (1.875) = 30 \text{ b/sec}$$

- Q) An analog signal band limited to 10KHz is quantized in 8 levels of a PCM system with probabilities of $1/4, 1/5, 1/5, 1/10, 1/10, 1/20, 1/20, 1/20$, respectively, find $H(x)$ & R ?

\rightarrow By Nyquist criterion

$$f_s \geq 2 f_m$$

$$f_s \geq 20 \text{ KHz}$$

$$H(x) = - \sum_{i=1}^{\infty} P(x_i) \log_2 P(x_i) \\ = - \left(\frac{1}{4} \frac{\log(1/4)}{\log 2} + \frac{2}{5} \frac{\log(1/5)}{\log 2} + \frac{2}{10} \frac{\log(1/10)}{\log 2} \right. \\ \left. + \frac{3}{20} \frac{\log(1/20)}{\log 2} \right) \\ = - (-0.5 + 0.928 - 0.66 - 0.604)$$

$$H(x) = 2.73 \text{ bits/symbol}$$

$$\gamma = 20K$$

$$R = 20K (2.73)$$

$$= \underline{54.7 \text{ K bit/sec}}$$

Q. A high resolution black & white TV picture consists of about 2×10^6 feature elements in 16 different brightness level. Pictures are repeated at the rate 32/sec. All picture elements are assumed to be independent and all levels have equal likelihood of occurrence. Calculate avg. rate of info. conveyed by this TV picture source?

$$\rightarrow m = 16 \quad p(x_i) = 1/16 \quad R = I(H(X)) \\ H(X) = - \sum_{i=1}^{16} \left(\frac{1}{16} \frac{\log(1/16)}{\log 2} \right) = 4 \quad = 2 \times 10^6 \times 32 \times 16 \\ \mu(x) = \underline{4 \text{ bits / symbols}}$$

Q. Given a telegraph source having 2 symbols (dot & -) the dot duration is 0.2 sec. The - duration is 3 times dot. If $p(\cdot) = 2 p(-)$ & the time b/w symbols is 0.2 sec calculate the information rate of telegraph source.

$$\rightarrow p(\cdot) = 2p(-) \\ p(\cdot) + p(-) = 3p(-) = 1 \quad \Rightarrow p(-) = \frac{1}{3}, \quad p(\cdot) = \frac{2}{3} \\ \text{From above two equation } p(-) = \frac{1}{3}, \quad p(\cdot) = \frac{2}{3} \\ H(X) = - \sum_{i=1}^{\infty} p(x_i) \log_2 p(x_i) \\ = - \left(\frac{2}{3} \frac{\log(1/3)}{\log 2} + \frac{1}{3} \frac{\log(2/3)}{\log 2} \right) \\ = - (-0.52 - 0.389) = \underline{0.91 \text{ bits / symbol.}}$$

$$R = I(H(X)) \quad t_{\cdot} = 0.2 \quad t_{-} = 0.6 \text{ sec}, \quad t_{\text{space}} = 0.2$$

TELEGRAPH $\rightarrow t_{\text{space}} =$

$$T_S = p(\cdot) t_{\cdot} + p(-) t_{-} + p(\cdot) t_{\text{space}}$$

$$T_S = 0.533 \text{ symbols.} \quad r = \frac{1}{T_S} = 1.675 \text{ symbols/sec}$$

$$R = 1.675(0.92) = \underline{1.525 \text{ bits/sec}}$$

Discrete memoryless channels (DMC)

Channel representation

$$[x] = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \xrightarrow{\text{channel}} p(y_i|x_i) \rightarrow [y] = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix}$$

A communication channel may be defined as a path or medium through which the symbols flow to the receiver end. DMC is a statistical model with a input x and a o/p y as shown above. During each unit of the time interval the channel accepts the i/p symbol from $[x]$ and in response it generates an o/p symbol $[y]$. This channel is said to be discrete when the alphabets of $[x]$ & $[y]$ are both finite also it is said to be memoryless & current o/p depends on only current i/p and not on any of the previous i/p. The possible i/p o/p path is indicated along with a conditional probability $p(y_i|x_i)$, where $p(y_i|x_i)$ is the conditional probability of obtaining o/p y_i given that the i/p is x_i . and is called a channel transition probability.

Channel matrix:

$$[p(y_i|x)] = \begin{bmatrix} p(y_1|x_1) & p(y_2|x_1) & \cdots & p(y_n|x_1) \\ p(y_1|x_2) & p(y_2|x_2) & \cdots & p(y_n|x_2) \\ \vdots & \vdots & \ddots & \vdots \\ p(y_1|x_n) & p(y_2|x_n) & \cdots & p(y_n|x_n) \end{bmatrix}$$

since each i/p to channel results in some o/p each row of the channel matrix must sum to unity

i.e., $\sum_{j=1}^n p(y_j|x_i) = 1$ for all i

Now if the i/p probability $p(x)$ are represented by row matrix then we have $[p(x)] = [p(x_1), p(x_2), \dots, p(x_n)]$ and also the o/p probabilities $[p(y)] = [p(y_1), p(y_2), \dots, p(y_n)]$

$$\text{then } [p(x)] = [p(x_1) \ p(x_2) \dots \ p(x_m)]$$

If $P(x)$ is represented as diagonal matrix, then we have

$$[p(x)] = \begin{bmatrix} p(x_1) & 0 & 0 & \dots & 0 \\ 0 & p(x_2) & 0 & \dots & 0 \\ \vdots & 0 & 0 & \dots & 0 \end{bmatrix}$$

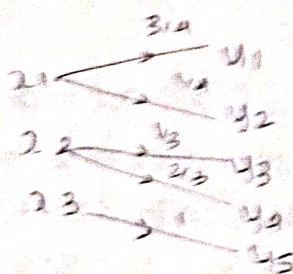
$$[p(x,y)] = p(x) \otimes p(y)$$

where i, j elements of matrix $p(x,y)$ has the form $p(x_i, y_j)$,
the matrix $p(x,y)$ is known as joint probability matrix of
 $\{p(x_i)\}$ and $\{p(y_j)\}$

acp types of channel:

loss less channel:

$$p(y|x) = \begin{bmatrix} 3/4 & 1/4 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 2/3 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

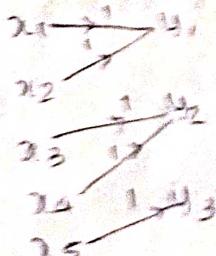


A channel is described by a channel matrix with only one non zero element in each column is called a loss less channel. It can be shown that in loss less channel no source information is lost in transmission.

Deterministic channel:

It is described by a channel matrix with only one non zero element in each row is called deterministic channel.

$$p(y|x) = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

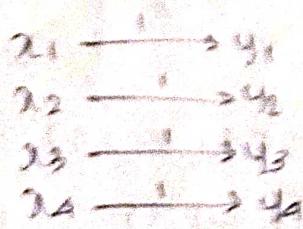


When a given source symbol is sent in deterministic channel it is clear that which channel symbol will be received.

Noiseless channel

A channel is called noiseless if it is both loss less and deterministic.

$$p(y|x) = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$



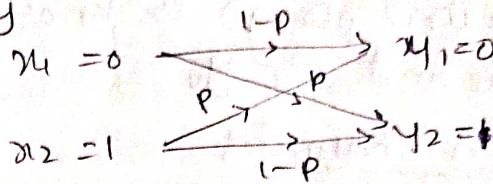
A noiseless channel matrix is shown in figure. The matrix has only one element in each row and each column & this element is unity. Note that the input & output symbols are same size i.e., $m=n$

Binary Symmetric channels:

It is defined by channel diagram

→ its channel matrix is given by

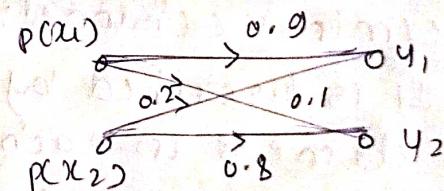
$$[P(Y|X)] = \begin{bmatrix} 1-p & p \\ p & 1-p \end{bmatrix}$$



It has 2 input $x_1=0$ and $x_2=1$ & 2 output $y_1=0, y_2=1$, this channel is symmetric bcz, the probability of receiving a 1 if a zero is sent is same as the probability of receiving a 0 if a one is sent. This common transition probability is denoted by p is shown above.

Q) Given a binary channel shown in fig. find channel matrix of the channel 2) find probability of $y_1 \& y_2$ when $P(x_1) = P(x_2) = 0.5$, 3) find the joint probabilities of $P(x_1, y_2) \& P(x_2, y_1)$ when $P(x_1) = P(x_2) = 0.5$

$$\Rightarrow P(Y|X) = \begin{bmatrix} P(Y_1|x_1) & P(Y_2|x_1) \\ P(Y_2|x_2) & P(Y_1|x_2) \end{bmatrix}$$



$$P(Y|X) = \begin{bmatrix} 0.9 & 0.1 \\ 0.2 & 0.8 \end{bmatrix}$$

$$P(Y) = P(X) P(Y|X)$$

$$= [P(x_1) \ P(x_2)] \begin{bmatrix} P(Y_1|x_1) & P(Y_2|x_1) \\ P(Y_1|x_2) & P(Y_2|x_2) \end{bmatrix}$$

$$= [0.5 \ 0.5] \begin{bmatrix} 0.9 & 0.1 \\ 0.2 & 0.8 \end{bmatrix}$$

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

$$\Rightarrow P(Y_1) = 0.55 \quad P(Y_2) = 0.45$$

$$P[X_1 Y_1] = P[X_1] P[Y_1 | X_1]$$

$$= \begin{bmatrix} 0.5 & 0 \\ 0 & 0.5 \end{bmatrix} \begin{bmatrix} 0.9 & 0.1 \\ 0.2 & 0.8 \end{bmatrix}$$

$$= \begin{bmatrix} 0.45 & 0.05 \\ 0.10 & 0.40 \end{bmatrix} \quad P[X_1, Y_1] = 0.05 \\ P[X_2, Y_1] = 0.10$$

(i) Two binary channels are connected in cascade as shown in the fig. Find the overall channel matrix of resultant channel and draw the equivalent channel diagram 2) find $P[Z_1]$ & $P[Z_2]$ when $P[X_1] = P[X_2] = 0.5$

$$P[Y] = [P(Y_1)] [P(Y_2 | X)]$$

$$= [P(Y_1)] [P(Z | Y)]$$

$$P[Z] = [P(Z_1)] P[(Y_1 | X)] P[(Z | Y)]$$

$$P[Z|x] = [P(Y_1 | X)] [P(Z | Y)]$$

$$P[Z|x] = \begin{bmatrix} 0.9 & 0.1 \\ 0.2 & 0.8 \end{bmatrix} \begin{bmatrix} 0.9 & 0.1 \\ 0.2 & 0.8 \end{bmatrix}$$

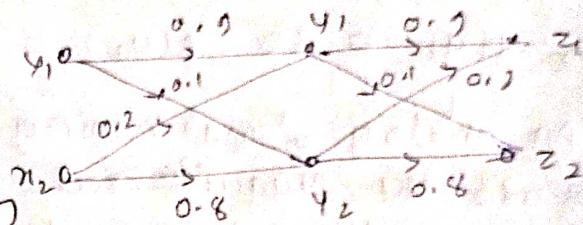
$$= \begin{bmatrix} 0.83 & 0.17 \\ 0.34 & 0.66 \end{bmatrix}$$

$$P[Z] = P[x] P[(Z|x)]$$

$$= [P(x_1) P(x_2)] \begin{bmatrix} 0.83 & 0.17 \\ 0.34 & 0.66 \end{bmatrix}$$

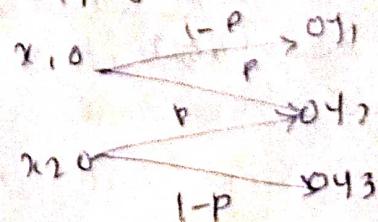
$$P[Z] = [0.585 \quad 0.415] \Rightarrow P(Z_1) = 0.585$$

$$P(Z_2) = 0.415$$



(ii) A channel has the following channel matrix. If the channel diagram 2) The source has equally likely a/p compute the probabilities associated with the channel a/p for $p = \text{channel}^2$.

$$\Rightarrow P(Y|x) = \begin{bmatrix} 1-p & p & 0 \\ 0 & p & 1-p \end{bmatrix}$$



Q) The binary erasure channel has 2 input $y_1 = 0, y_2 = 1$ & 3 output $y_1 = 0, y_2 = c, y_3 = 1$, where c indicates an erasure. This means that the outputs in doubt & hence should be erased. $\Rightarrow p(y_1) = p(y_2) = p(y_3) = \frac{1}{3}$ substitute values

$$p(y) = [p(y_1)] [p(y_2)]$$

$$= [0.5 \quad 0.5] \begin{bmatrix} 0.8 & 0.2 & 0 \\ 0.2 & 0.8 & 0 \\ 0 & 0 & 0.2 \end{bmatrix}$$

$$= [0.40 \quad 0.20 \quad 0.10]$$

$$\rightarrow p(y_1) = 0.40 \quad p(y_2) = 0.20 \quad p(y_3) = 0.10$$

Q) An analog signal having 4 kHz bandwidth is sampled at 10 times the Nyquist rate & each sample is quantized into one of equally likely levels. Assume that successive samples are statistically independent. If Rq is the OIP of the source, it can be transmitted without error over an additive white Gaussian channel with $BW = 10\text{kHz}$ & $SNR = 20\text{dB}$. Find the SNR required for error free transmission for path 1. Q) find the BW required for a AWGN channel for error free transmission of OIP of this source if $SNR = 20\text{dB}$

$$\rightarrow f_m = 4\text{kHz} \quad f_s = 2f_m$$

$$\text{Nyquist rate} \rightarrow r = 2f_m (1.25) = 10\text{kHz}$$

$$r = \underline{\underline{10000 \text{ samples/sec}}}$$

$$R = r f_l(x)$$

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

$$C = 10^4 \log_2 [1 + 100]$$

$$C = \frac{10^4 \log(101)}{100^2} \quad C = \underline{\underline{66.6 \times 10^3 \text{ bits/sec}}}$$

∴ SNR ratio can be found by

$$C = 10^4 \log_2 (1 + \frac{S}{N}) \geq 8 \times 8 \times 10^3$$

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

$$1 + \frac{S}{N} \geq 2^8 = 256 \Rightarrow \frac{S}{N} \geq 255 = \underline{\underline{24.1 \text{ dB}}}$$

$I = ? \text{ Hertz}$

$$I = 10^4 \times 8 = \underline{\underline{80 \text{ bits}}}$$

$$H(x) = \log_2 (P(x)) = \log_2 256 = 8 \text{ bits/samples.}$$

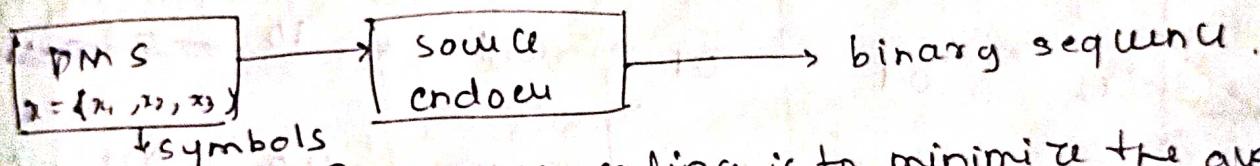
$$p(x_i) = \frac{1}{256}$$

The required BW can be found by

$$(= B \log_{10} [1 + 100 \times 8 \times 10^3] \Rightarrow \underline{\underline{B \geq 1.2 \text{ kHz}}}$$

Source Coding:

To conversion of o/p of DMS into a sequence of binary symbols (binary code word) is called source coding.



An objective of source coding is to minimize the avg. bit rate required for representation of the source by reducing the information source.

Code length and efficiency:

Let x be a DMS with finite entropy $H(x)$ and an alpha set x_1, x_2, \dots, x_m with corresponding probabilities of occurrence $p(x_i)$ for $i = 1, 2, \dots, m$, the avg. code word length

$$L = \sum_{i=1}^m p(x_i) n_i$$

$n_i \rightarrow$ length of code word

The parameter L avg. no. of bits per source symbol used in the source coding process also the code efficiency is defined as

$$n = \frac{L_{\min}}{L}$$

$L_{\min} \rightarrow$ the minimum possible value of L

When η approaches unity, the code is said to be efficient. The code redundancy, γ , is defined as

$$\gamma = 1 - \eta$$

source coding theorem:

It states that, for a DMs (X) with entropy ($H(x)$) the avg. code word length L per symbol bounded by $L \geq H(x)$ and further L can be made as close to $H(x)$ as desired for some suitably chosen code, thus with $L_{\min} = H(x)$, the code efficiency can be rewritten as

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

Classification of Codes :-

x_i	Code 1	Code 2	Code 3	Code 4	Code 5	Code 6
x_1	00	00	0	0	0	1
x_2	01	01	1	10	01	01
x_3	00	10	00	10	011	001
x_4	10	11	11	11	0111	0001

Fixed Length Code:

It is one whose code length is fixed. Code 1 & Code 2 of this table are fixed length with ($L = 2$).

Variable Length Code:

Variable length code is one whose code word length is not fixed. Code (3, 4, 5, 6).

Distinct codes:

The code is distinct if each code word is distinct from other code words. Code (2, 3, 4, 5, 6)

prefix free codes:

A code in which no code word can be formed by adding code word symbols to another code word is called a prefix free code. Thus, in a prefix free code, no code word is a prefix of another. Code $(2, 1, 6)$

uniquely decodable codes:

A distinct code uniquely decodable if the original source sequence can be reconstructed perfectly from the encoded binary sequence. Note that code 3 of the table is not a uniquely decodable code for eg; the binary sequence 1001 may correspond to source sequences x_2, x_3, x_2 , or x_2, x_1, x_1, x_2 . A sufficient condition to insure a code is uniquely decodable is that no code word is a prefix of another, thus the prefix free code $(2, 1, 6)$ are uniquely decodable codes.

instantaneous codes:

An uniquely decodable code is called instantaneous code, the end of any code word is recognizable without examining subsequent code symbols. so the prefix free codes are sometimes known as instantaneous codes.

optimal codes:

A code is said to be optimal if it is instantaneous and has a minimum avg. L for a given source with a given probability for the source symbols.

length kraft inequality:

Let X be a DMS with alphabet $x_i = h \mid i=1, 2, \dots, m$. Assume that the length of the assigned binary code word corresponding to $[x_i]$ is n_i . A necessary & sufficient conditions for existence of a instantaneous binary code is

$$K = \sum_{i=1}^m 2^{-n_i} \leq 1 \rightarrow \text{kraft inequality}$$

- i) consider a PMS(X) with symbols x_i ($i=1, 2, 3, 4$)
ii) show that all the codis (except code V) satisfy the Kraft inequality
iii) show that codis(A & D) are uniquely decodable but codis(B)
are not uniquely decodable.

\rightarrow

$$K = \sum_{i=1}^m 2^{-n_i}$$

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

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

$$+ \frac{1}{8} + \frac{1}{8} + \frac{1}{8} = 1$$

x_i	code A	code B	code C	code D
x_1	00	0	0	0
x_2	01	10	11	100
x_3	10	11	100	110
x_4	11	110	100	111
	2	1223	1233	133
	$K = 1$	1.12	1	0.8

code A & D are prefix free codes, uniquely decodable.

Entropy Coding :-

The design of a variable length code such that its avg. code word length approaches the entropy of PM is often referred to as entropy coding.

Shanon Fano Coding:

An efficient code can be obtained by following simple procedure known as shanon fano algorithm

- 1) list the source symbols in order of decreasing probability.

- 2) partition the set into two sets that are as close to equi-probable as possible, and assign '0' to upper set and '1' to lower set.

- 3) continue this process each time partitioning the sets with as nearly equal probabilities as possible until further partitioning is not possible.

Q) A PMS x_i has 6 symbols $x_1, x_2, x_3, \dots, x_6$ with probability $p(x_1) = 0.30, p(x_2) = 0.25, p(x_3) = 0.20, p(x_4) = 0.12, p(x_5) = 0.08, p(x_6) = 0.05$. construct a shanon fano code for x , find the efficiency and redundancy of the code?

x_i	$p(x_i)$	Step 1(.)	Step 2(.)	Step 3(.)	Step 4(.)
x_1	0.30	0	0	0	0
x_2	0.25	0	1	0	0
x_3	0.20	1	0	0	0
x_4	0.12	1	1	0	0
code 0	x_5	1	1	1	0
0	x_6	1	1	1	1
100	Code	n			
110	x_1	00	2		
111	x_2	01	2		
1111	x_3	10	2		
11111	x_4	110	3		
111111	x_5	1110	4		
1111111	x_6	1111	4		

$$n = \frac{H(x)}{2}$$

$$H(x) = -\sum_{n=1}^{\infty} p(x_n) \log_2 p(x_n)$$

$$\begin{aligned} H(x) &= 0.30 \log_2 0.30 + 0.25 \log_2 0.25 \\ &\quad + 0.2 \log_2 0.2 + 0.12 \log_2 0.12 \\ &\quad + 0.08 \log_2 0.08 + 0.05 \log_2 0.05 \end{aligned}$$

$$H(x) = +2.36$$

$$\begin{aligned} L &= \sum_{i=1}^6 p(x_i) n_i = 0.30(2) + 0.25(2) + 0.20(2) \\ &\quad + 0.12(3) + 0.08(4) + 0.05(4) \\ &= 0.60 + 0.5 + 0.4 + 0.36 + 0.32 + 0.20 \\ &= 2.10 + 0.76 + 0.52 \\ &= 1.86 + 0.52 = \underline{\underline{2.38}} \end{aligned}$$

$$\eta = \frac{2.36}{2.38} = 0.991 \times 100\% = \underline{\underline{99.16\%}}$$

Q) A DMS x has 5 symbols x_1, x_2, \dots, x_5 with $p(x_1) = 0.4$, $p(x_2) = 0.19$, $p(x_3) = 0.16$, $p(x_4) = 0.15$, $p(x_5) = 0.1$, construct a shanon fano code & calculate the efficiency of code

x_i	$p(x_i)$	Step 1	Step 2	Step 3	Code	n_i
x_1	0.4	0	0		00	2
x_2	0.19	0	1		01	2
x_3	0.16	1	0	*	10	2
x_4	0.15	1	1	0	110	3
x_5	0.1	1	1	1	111	3

$$H(2) = - \sum_{n=1}^S p(x_i) \log_2 p(x_i)$$

$$= -(0.1 \log_2 0.1 + 0.19 \log_2 0.19 + 0.16 \log_2 0.16 \\ + 0.15 \log_2 0.15 + 0.1 \log_2 0.1)$$

$$= \underline{2.14}$$

$$L = \sum_{n=1}^S p(x_i) n i$$

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

$$= 0.8 + 0.38 + 0.32 + 0.45 + 0.3$$

$$= \underline{\underline{2.25}}$$

$$n = \frac{2.14}{2.25} = \underline{\underline{95.11\%}}$$

Huffman Coding:

In general huffman encoding results in an optimum code thus it is the code that has the highest efficiency.
 0) list the source symbols in order of decreasing probability.
 2) combine the probabilities of 2 symbols having the lowest probabilities & reorder the resultant probabilities. This step is known as reduction (1). The same procedure is repeated until there are 2 order probabilities remaining.

3) start encoding with the last reduction which consists of exactly two order probabilities. Assign '0' to first set & '1' to the lower set.

4) now go back & assign 0 & 1 to the second digit for the two probabilities that were combined in previous reduction step.

5) keep regressing this way until first column is reached.

Q) $p(x_i)$

							Code	n _i
x ₁	0.30	0.30	0.30	0.45	0.55	0	00	2
x ₂	0.25	0.25	0.25	0.30	0.45	1	02	2
x ₃	0.20	0.20	0.25	0.25	1		11	2
x ₄	0.12	0.13	0	0.20	1		101	3
x ₅	0.08	0	0.12	1			1000	4
x ₆	0.05	1					1001	4

$$H(X) = 2.36, L = 2.38$$

$$\eta = \underline{99.164}$$

Q) A DMS X has 5 symbols x_1, \dots, x_5 , $p(x_1) = 0.4$, $p(x_2) = 0.19$, $(x_3) = 0.16$, $(x_4) = 0.15$, $(x_5) = 0.1$.

→ $x_i \ p(x_i)$

				Code	n _i
x ₁	0.4	0.4	0.4	1	1
x ₂	0.19	0.25	0.35	000	3
x ₃	0.16	0.19	0.25	001	3
x ₄	0.15	0.16	0.16	010	3
x ₅	0.1	1		011	3

$$H(X) = 2.14$$

$$L = \sum p(x_i) n_i$$

$$= 1(0.4) + 3(0.19) + 3(0.16) + 3(0.15) + 3(0.1)$$

$$= 0.4 + 0.57 + 0.48 + 0.45 + 0.3$$

$$= 2.35$$

$$\eta = \frac{2.14}{2.35} = \underline{\underline{97.37}}$$