

Information Theory and Channel Capacity

Module 1

GATE: Section 7

part of GATE Syllabus (ECE)
Section 7, Communication

Meaning of Information: Message or Intelligence.

Source → Which produces message or intelligence



Information Source: 1) Analog information sources: microphone, TV camera etc.

2) Digital information sources: Fax etc, teletype etc.
(Discrete) (Encoded)

→ Discrete information sources are characterized by: a) source alphabet, b) symbol rate c) source alphabet probabilities and d) probabilistic dependence of symbols in a sequence.

Symbol rate: Rate at which teletype produces characters
e.g., 100 chars./sec. ⇒ symbol rate: 10 symbols/sec.

Source alphabets & their probabilities:

$$S = \{s_1, s_2, s_3, \dots, s_q\} \text{ \& } P = \{p_1, p_2, p_3, \dots, p_q\}$$

i.e., $s_1 \rightarrow p_1; s_2 \rightarrow p_2; s_3 \rightarrow p_3$ so on...

$$\text{Then } p_1 + p_2 + p_3 + \dots + p_q = 1$$

$$\text{or } \sum_{i=1}^q p_i = 1.$$

C. E. Shannon → Father of Information Theory → presented a paper on
Mathematical Theory of Communication → Theoretical bounds for the performances of Communication Systems.Measure of Information:Let $S = \{s_1, s_2, s_3, \dots, s_q\}$ with probabilities $P = \{p_1, p_2, \dots, p_q\}$

Then Self information or Amount of information

$$I_k = \log \frac{1}{p_k}$$

rate: log base is '2'. i.e., (Bits) (common)

Also, when base is '10' units → Hartleys
or Decis

and, when base is 'e' units → NATs

or In general for the base of 'r' units are called r-ary units.

Example: On a particular day during winter season in

1. Sun will rise in east on the day of trip $P=1$; $I=0$ no surprise!
(Certain)
2. It will be a very cold day $P \neq 0$; $I \neq 0$
3. There will be snowfall on that particular day in $P=0$; $I \rightarrow \text{Max}$
↑ full of surprise!
Uncertain

⇒ The only equation which satisfies 1 to 3 is:

$$\Rightarrow I = \log \frac{1}{P}$$

example 1: The binary symbols 0 & 1 are transmitted with probabilities $1/4$ & $3/4$ respectively. Find 'I' in each case:

→ Symbol '0' with $P_0 = 1/4$

$$I_0 = \log_2 \frac{1}{P_0} = \log_2 \frac{1}{1/4} = 2$$

$$\left. \begin{array}{l} \log_2 2 = 1 \\ \log_2 m^n = n \log_2 m \end{array} \right\}$$

|||

$$I_1 = \log_2 \frac{1}{P_1} = \log_2 \left(\frac{1}{3/4} \right) = 0.415 \text{ bits} \Rightarrow \text{less info for high 'P'}$$

Why 'log':
Info can't be -ve.
 $I=0$ for sure event (lower bound) ($P=1$)
 $I>0$ for $P \neq 0$
 $I \rightarrow \text{max}$ for least-probable events ($P=0$).

Property: When independent symbols are transmitted, the total self information must be equal to the sum of individual self informations.

proof: Let us consider two independent symbols s_k and s_l are transmitted with probabilities P_k and P_l .

$$\rightarrow I_{kl} = \log \frac{1}{P(s_k \text{ and } s_l)} = \log \frac{1}{(s_k \text{ and } s_l)}$$

$$= \log \frac{1}{P(s_k) \cdot P(s_l)} = \log \frac{1}{P_k P_l} = \log \frac{1}{P_k} + \log \frac{1}{P_l} = I_k + I_l$$

$$\Rightarrow I_{kl} = I_k + I_l$$

→ Total self information is equal to the sum of individual self informations.

Zero-Memory Source: Statistically independent symbols i.e., no connection between the symbols i.e., source has no memory. \Rightarrow Zero memory or Memoryless.

Average Information (H) OR Entropy: (GATE Syllabus)

\triangleq Average self information = I_{total}/L

$$\text{Where } I_{\text{tot}} = L \sum_{i=1}^q P_i \log_2 \frac{1}{P_i} \text{ bits}$$

\Rightarrow Average self information is also called "ENTROPY".

$$\therefore \text{Entropy of source 'S' is } H(S) = \sum_{i=1}^q P_i \log_2 \frac{1}{P_i}$$

Other names of $H(S)$: Average uncertainty or Average amount of surprise.

(1) Example: Find $H(S)$ for $S = \{s_1, s_2\}$ with $P = \left\{ \frac{1}{256}, \frac{255}{256} \right\}$

$$\rightarrow H(S) = \frac{1}{256} \log_2 \frac{1}{1/256} + \frac{255}{256} \log_2 \frac{1}{255/256}$$

$$= 0.037 \text{ bits/symbol}$$

Observation: Avg. uncertainty is very small as occurrence of s_1 & s_2 can be predicted easily (prob. difference)

(2) Example: Repeat (1) for $S = \{s_3, s_4\}$ with $P = \left\{ \frac{1}{2}, \frac{1}{2} \right\}$

$$\rightarrow H(S) = \frac{1}{2} \log_2 \frac{1}{1/2} + \frac{1}{2} \log_2 \frac{1}{1/2} \Rightarrow \frac{2}{2} \log_2 2 = \frac{2}{2} = 1 \text{ bits/symbol}$$

Observation: Uncertainty (avg.) is maximum: Impossible to predict.

Information Rate: Let r_s be the rate at which source emits the symbols
Then Average source information rate ' R_s ' in bit/sec is the product of $H(S)$ & r_s : i.e., Average information content per symbol & message symbol rate

$$\therefore R_s = r_s \cdot H(S) \text{ bits/sec}$$

Example: $S = \{s_1, s_2, s_3\}$ with $P = \left\{ \frac{1}{2}, \frac{1}{4}, \frac{1}{4} \right\}$

Find a) Self Information

b) Entropy of Source

$$\begin{aligned}
 \text{Solution: } H(X) &= P_1 I_1 + P_2 I_2 + P_3 I_3 \\
 &= P_1 \log_2 \frac{1}{P_1} + P_2 \log_2 \frac{1}{P_2} + P_3 \log_2 \frac{1}{P_3} = \sum_{i=1}^3 P_i \log_2 \frac{1}{P_i} \\
 &= \frac{1}{2} \log_2 \left(\frac{1}{1/2}\right) + \frac{1}{4} \log_2 \left(\frac{1}{1/4}\right) + \frac{1}{4} \log_2 \left(\frac{1}{1/4}\right) \\
 &= \frac{1}{2} + \frac{1}{2} + \frac{1}{2} = 1.5 \text{ bits/symbol}
 \end{aligned}$$

Relationship among the parameters: Bits, Nats, and Hartleys
 → Difference in 'log' base

$$\begin{aligned}
 \text{i.e., } I &= \log_{10} \frac{1}{P} \text{ Hartleys} && \text{base: } 10 \\
 &= \log_e \frac{1}{P} \text{ Nats} && \text{base: } e \\
 &= \log_2 \frac{1}{P} \text{ bits} && \text{base: } 2
 \end{aligned}$$

a) Relationship between Hartley & Nats

$$\begin{aligned}
 I &= \log_{10} \frac{1}{P} \text{ Hartleys} \\
 \therefore 1 \text{ Hartley} &= \frac{I}{\log_{10} \frac{1}{P}} = \frac{\log_e \frac{1}{P}}{\log_{10} \frac{1}{P}} = \frac{-\log_e P}{-\log_{10} P}
 \end{aligned}$$

$$\rightarrow \frac{\frac{1}{\log_e P}}{\frac{1}{\log_{10} P}} = \frac{\log_{10} P}{\log_e P} \text{ Nats}$$

$$\left\{ \begin{aligned} \log_a b &= \frac{1}{\log_b a} \end{aligned} \right.$$

$$\begin{aligned}
 &= \frac{\log_{10} 10}{\log_e 10} = \log_e 10 = 2.303 \text{ Nats} \\
 &\frac{\log_e 10}{\log_e e} = \log_e 10 = 2.303 \text{ Nats}
 \end{aligned}$$

$$\log_n^m = \frac{\log_e^m}{\log_e^n}$$

$$\log_e^e = \log_2^2 = 1$$

$$\text{Similarly: } 1 \text{ Hartley} = \log_2^{10} \frac{\text{bits}}{\text{Nats}} = 2.303 \text{ Nats}$$

$$\Rightarrow 1 \text{ Hartley} = 3.32 \text{ bits} = 2.303 \text{ Nats}$$

$$\text{And, } 1 \text{ nat} = \log_2^e \text{ bits}$$

$$\therefore 1 \text{ nat} = \frac{1}{\log_e 2} = 1.443 \text{ bits}$$

Example: A fair coin is tossed repeatedly. Let-

A = event of getting 3 heads out of 5 trials

B = event of getting 5 heads out of 8 trials

Which event conveys more information? Support answer by numerical computation of respective amounts of information.

Soln: Let X be the R.V. = no of heads

= Binomial R.V. (since tossing a coin results in only 2 outcomes)

$n \rightarrow$ no. of trials

And, $p \rightarrow$ prob. of getting H,

$1-p \rightarrow$ " " " T

$p=q=0.5$ Equiprobable & mutually exclusive

Then $\Delta P(X=x) = {}^n C_x p^x q^{n-x}$

$${}^n C_r = \frac{n!}{r!(n-r)!}$$

e.g. ${}^4 C_3 = \frac{4!}{3!(4-3)!}$

Also $0! = 1! = 1$.

$\Rightarrow P(A) = P(X=3) = {}^5 C_3 p^3 q^{5-3} = 0.3125$

$\Rightarrow P(B) = P(X=5) = {}^8 C_5 p^5 q^{8-5} = 0.21875$

$\rightarrow I_A = \log_2 \frac{1}{P(A)} = \log_2 \frac{1}{0.3125} = 1.678 \text{ bits}$

$\Rightarrow I_B = \log_2 \frac{1}{P(B)} = \log_2 \frac{1}{0.21875} = 2.193 \text{ bits}$

$\therefore I_B > I_A$ event B conveys more information!

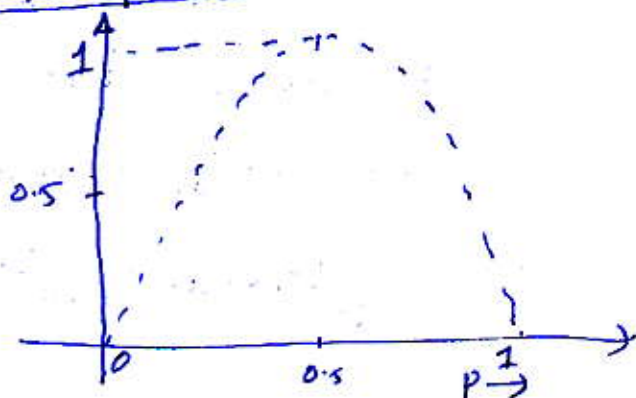
Example: A binary source is emitting an independent sequence of 0s and 1s with probabilities p and $1-p$ respectively. plot the entropy of the source versus p .

\rightarrow Entropy $H(s) = \sum_{i=1}^2 p_i \log_2 \frac{1}{p_i} = p \log_2 \frac{1}{p} + (1-p) \log_2 \frac{1}{1-p}$

p	0	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0
$H(s)$	0	0.469	0.722	0.881	0.991	1	0.991	0.881	0.722	0.469	0

$\rightarrow H(s)$ is max when $p=0.5$
i.e., Max uncertainty

$\rightarrow H(s)$ is symmetric



Properties of Entropy:

For a source alphabet- $S = \{s_1, s_2, \dots, s_q\}$ with $P = \{p_1, p_2, \dots, p_q\}$
where $q =$ number of source symbols, Then

$$H(S) = \sum_{i=1}^q p_i \log \frac{1}{p_i} = \sum_{i=1}^q p_i I_i \quad \text{bits/symbol} \quad \text{Entropy or Avg. Info? (self)}$$

Following properties can be noted:

a) The entropy function is continuous for every independent-variable p_k in the interval $(0, 1)$ i.e., $p_k \uparrow; H(S) \downarrow$

b) The entropy function is a symmetrical function of its arguments.

$$\text{i.e., } H(p_k, (1-p_k)) = H((1-p_k), p_k) \quad \forall k; k=1, 2, 3, \dots, q$$

the value of $H(S)$ remains same irrespective of location of probabilities

$$\text{e.g., } P_A = \{p_1, p_2, p_3\}, P_B = \{p_2, p_3, p_1\}, \text{ and } P_C = \{p_3, p_1, p_2\}$$

Then $H(S_A) = H(S_B) = H(S_C)$; S_A, S_B, S_C are sources.

c) External property: Let $S = \{s_1, s_2, s_3, \dots, s_q\}$ with $P = \{p_1, p_2, \dots, p_q\}$

The entropy of S is given by

$$H(S) = \sum_{i=1}^q p_i \log \frac{1}{p_i} \quad \text{and} \quad \sum_{i=1}^q p_i = 1$$

Then upper & lower bound of $H(S)$ are

\Rightarrow It is obvious that lower bound for $H(S) = 0$.

Upper bound: $\log q - H(S) = \left(\sum_{i=1}^q p_i \right) \log q - \sum_{i=1}^q p_i \log \frac{1}{p_i} \quad (1) \quad \because \sum_{i=1}^q p_i = 1.$

$$= \sum_{i=1}^q p_i \left[\log q - \log \frac{1}{p_i} \right]$$

$$= \sum_{i=1}^q p_i \log 2 p_i$$

$$= \sum_{i=1}^q p_i \frac{\log_e 2 p_i}{\log_e 2}$$

$$\log \left(\frac{m}{n} \right) = \log m - \log n$$

$$\log_2^m = \frac{\log_e m}{\log_e 2}$$

$$\therefore \log q - H(S) = \log_2^e \sum_{i=1}^q p_i \ln 2 p_i \quad (2)$$

Using log inequality $y = x - 1$ & $y = \ln x \Rightarrow \ln x \leq x - 1$
equality only at $x = 1$

$$\text{Let } x = \frac{1}{2 p_i} \rightarrow \ln 2 p_i \geq 1 - \frac{1}{2 p_i} \quad \text{or } \ln \frac{1}{x} \geq 1 - x$$

$$\Rightarrow \sum_{i=1}^q p_i \ln 2 p_i \geq \sum_{i=1}^q p_i \left(1 - \frac{1}{2 p_i} \right) \quad (3) \quad \because \text{Multiplying both sides by } \sum_{i=1}^q p_i$$

Multiplying (3) by $\log_2 e$ both sides we get-

$$\log_2 e \sum_{i=1}^q P_i \ln q P_i > \log_2 e \left[\sum_{i=1}^q P_i - \sum_{i=1}^q \frac{1}{q} \right] \quad (4)$$

RHS is always zero & LHS is $\log_2 q - H(S)$

$$\text{i.e., } \log_2 q - H(S) > 0$$

$$\text{or } H(S) \leq \log_2 q$$

The equality sign holds good when $P_i - \frac{1}{q} = 0 \quad \forall i$

$$\text{i.e., } P_i = \frac{1}{q} \quad \forall i \quad i = 1, 2, 3, \dots, q.$$

$$\therefore H(S)_{\max} = \log_2 q \text{ bits/symbol.} \quad q \rightarrow \text{number of symbols.}$$

\therefore Entropy attains max when all symbols are equiprobable.

*) Example: $S = \{s_1, s_2, s_3, s_4\}$ with $P = \{1/4, 1/4, 1/4, 1/4\}$ $P_i = 1/4$ for $\forall i$

$$\Rightarrow H(S) = H(S)_{\max} = \log_2 4 = 2 \text{ bits/symbol}$$

$$\text{Alternatively, } 4 \times \frac{1}{4} \log_2 \frac{1}{1/4} = 2 \text{ bits/symbol.}$$

(d) Property of Additivity: Suppose splitting the symbols S_q into $S_{q_1}, S_{q_2}, \dots, S_{q_n}$ with probabilities $P_{q_1}, P_{q_2}, \dots, P_{q_n}$ such that-

$$P_q = P_{q_1} + P_{q_2} + P_{q_3} + \dots + P_{q_n} = \sum_{j=1}^n P_{q_j} \quad (1)$$

Then, the splitted symbol entropy is

$$H'(S) = H(P_1, P_2, P_3, \dots, P_{q-1}, P_{q_1}, P_{q_2}, \dots, P_{q_n}) \text{ such that}$$

$$= \sum_{i=1}^{q-1} P_i \log \frac{1}{P_i} + \sum_{j=1}^n P_{q_j} \log \frac{1}{P_{q_j}}$$

$$= \sum_{i=1}^q P_i \log \frac{1}{P_i} - P_q \log \frac{1}{P_q} + \sum_{j=1}^n P_{q_j} \log \frac{1}{P_{q_j}} \quad (2)$$

$$= \sum_{i=1}^q P_i \log \frac{1}{P_i} - \sum_{j=1}^n P_{q_j} \log \frac{1}{P_q} + \sum_{j=1}^n P_{q_j} \log \frac{1}{P_{q_j}}$$

$$= H(S) - \sum_{j=1}^n P_{q_j} \left(\log \frac{1}{P_{q_j}} - \log \frac{1}{P_q} \right)$$

After arranging & simplifying

$$\therefore H'(S) = H(S) + \text{a positive quantity} \quad \because P_{q_j} \leq P_q \quad \forall j$$

$$\therefore H'(S) \geq H(S)$$

"Partitioning the symbols into subsymbols will not reduce entropy"

(e) Source Efficiency: Ratio of entropy to max entropy and is given by $\eta_s = \frac{H(S)}{H(S)_{max}}$ and the source redundancy denoted by

$$R_{\eta_s} \text{ is given by } R_{\eta_s} = 1 - \eta_s$$

Note: η_s and R_{η_s} can also be expressed in %.

Example: Verify the rule of additivity:

$$S = \{s_1, s_2, s_3, s_4\} \text{ with } P = \{1/2, 1/3, 1/12, 1/12\} = (P_1, P_2, P_3, P_4) \text{ say}$$

$$\rightarrow H(S) = 1.625 \text{ bits/symbol}$$

Using property of additivity

$$H'(S) = P_1 \log \frac{1}{P_1} + (1-P_1) \log \frac{1}{1-P_1} + 1-P_1 \left\{ \frac{P_2}{1-P_1} \log \frac{1-P_1}{P_2} + \frac{P_3}{1-P_1} \log \frac{1-P_1}{P_3} + \frac{P_4}{1-P_1} \log \frac{1-P_1}{P_4} \right\}$$

$$\therefore H'(S) = \frac{1}{2} \log 2 + \frac{1}{2} \log 2 + \frac{1}{2} \left\{ \frac{1/3}{1/2} \log \left(\frac{1/2}{1/3} \right) + \frac{1/12}{1/2} \log \frac{1/2}{1/12} + \frac{1/12}{1/2} \log \frac{1/2}{1/12} \right\}$$

$$H'(S) = 1.625 = H(S)$$

Example: A black and white TV picture consists of 525 lines of picture information. Assume that each line consists of 525 picture elements & that each element can have 256 brightness levels. Picture are repeated at the rate of 30 frames/sec. Calculate the average rate of information conveyed by a TV set to a viewer

Solution: No. of lines = 525; picture elements/line = 525

$$\therefore \text{Total no. of elements} = 525 \times 525$$

Also, each element can have 256 brightness levels.
Then each frame can have brightness levels of $256^{525 \times 525}$

$$\therefore \text{Max. Avg. Inf}^n = H(S)_{max} = \log_{256} 256^{525 \times 525} = 22.05 \times 10^5 \text{ bits/frame.}$$

$$\therefore \text{Avg. Inf}^n \text{ rate } R_s = \rho \gamma_s \cdot H(S)_{max} = \gamma_s I_{max}(R_s)$$

$$= 30 \times 22.05 \times 10^5$$

$$R_s = 66.15 \times 10^6 \text{ bits/sec}$$

⑤

Suppose that S_1 and S_2 are two zero-memory sources with probabilities p_1, p_2, \dots, p_n for source S_1 and q_1, q_2, \dots, q_n for source S_2 . Show that the entropy of source S_1

$$H(S_1) \leq \sum_{k=1}^n p_k \log \frac{1}{q_k}$$

Solution: Given S_1 & S_2 are zero memory sources.

$$\therefore H(S_1) = \sum_{k=1}^n p_k \log \frac{1}{p_k} \quad \text{--- (1)}$$

$$\& \sum_{k=1}^n p_k = 1 \quad (\text{By defn of probability})$$

$$\text{why } H(S_2) = \sum_{k=1}^n q_k \log \frac{1}{q_k} \& \sum_{k=1}^n q_k = 1. \quad \text{--- (2)}$$

$$\text{Consider: } H(S_1) - \sum_{k=1}^n p_k \log \frac{1}{q_k} = \sum_{k=1}^n p_k \log \frac{1}{p_k} - \sum_{k=1}^n p_k \log \frac{1}{q_k}$$

$$= \sum_{k=1}^n p_k \left(\log \frac{1}{p_k} - \log \frac{1}{q_k} \right) = \sum_{k=1}^n p_k \cdot \log_2 \left(\frac{q_k}{p_k} \right) \quad \left\{ \text{log rule} \right.$$

$$= \sum_{k=1}^n p_k \left[\frac{\log_e \left(\frac{q_k}{p_k} \right)}{\log_e 2} \right] = \log_e 2 \sum_{k=1}^n p_k \ln \left(\frac{q_k}{p_k} \right) \quad \left\{ \begin{array}{l} \log_2 e = \frac{\log_e e}{\log_e 2} = \frac{1}{\log_e 2} \\ \log_e 2 = \frac{1}{\log_2 e} \end{array} \right.$$

$$\text{W.K.T } \ln \frac{1}{x} \geq 1-x$$

$$-\ln x \geq 1-x$$

Removing -ve sign i.e., $\ln x \leq x-1$

$$\text{let } x = \frac{q_k}{p_k} \therefore \ln \left(\frac{q_k}{p_k} \right) \leq \frac{q_k}{p_k} - 1$$

Multiply by p_k taking summation for $\forall k (k=1, 2, \dots, n)$ & then multiply - ing by $\log_e 2$ on both sides we get -

$$\log_e 2 \sum_{k=1}^n p_k \ln \left(\frac{q_k}{p_k} \right) \leq \log_e 2 \sum_{k=1}^n p_k \left(\frac{q_k}{p_k} - 1 \right)$$

$$= H(S_1) - \sum_{k=1}^n p_k \ln \left(\frac{q_k}{p_k} \right) \leq \log_e 2 \underbrace{\sum_{k=1}^n (q_k - p_k)}_{=0} = 0$$

$$\therefore H(S_1) \leq \sum_{k=1}^n p_k \log \frac{1}{q_k}$$

Imp step

log inequality

Example: A discrete message source "S" emits two independent-symbols X and Y with probabilities 0.55 and 0.45 respectively. Calculate the efficiency of the source and its redundancy.

$$P_x = p(X) = 0.55, \quad P_y = p(Y) = 0.45 \quad (X, Y) = (s_1, s_2)$$

$$\therefore \text{Entropy } H(S) = \sum_{i=1}^n P_i \log_2 \frac{1}{P_i}$$

$$\therefore H(S) = P_x \log_2 \frac{1}{P_x} + P_y \log_2 \frac{1}{P_y}$$

$$= 0.55 \log_2 \frac{1}{0.55} + 0.45 \log_2 \frac{1}{0.45}$$

$$= 0.9928 \text{ bit/symbol}$$

$$\eta_s = \frac{H(S)}{H(S)_{\max}} = \frac{0.9928}{1} = 99.28\%$$

And, the source redundancy is given by

$$R_{ns} = 1 - \eta_s = 1 - 0.9928 = 0.0072$$

$$\therefore R_{ns} = 0.72\%$$

Example: In a facsimile transmission of picture, there are about 2.25×10^6 pixels/frame. For a good reproduction 12 brightness levels are necessary. Assume all these levels are equally likely to occur. Find the rate of information if one picture is to be transmitted every 3 minutes. What is the source efficiency of this fax transmitter.

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Solution: Total number of pixels in one frame = 2.25×10^6
Number of brightness levels = 12

$$\therefore \text{Total no. of different frames possible} = \frac{2.25 \times 10^6}{12}$$

Since all levels are equally likely to occur, the net max info content per frame is

$$I = H(S)_{\max} = \log_2 12 = \log_2 (12)$$

$$= 2.25 \times 10^6 \log_2 12 \quad \therefore I = 8.066 \times 10^6 \text{ bits/frame} = H(S)_{\max}$$

Given that one picture is transmitted in 3 minutes

Therefore, the rate of transmission is given by

$$r_s = \frac{1}{3 \text{ minutes}} = \frac{1}{3 \times 60} \text{ Pictures/sec}$$

∴ The average rate of information is given by

$$R_s = r_s I = \frac{1}{3 \times 60} \times 8.066 \times 10^6$$

$$R_s = 44812 \text{ bits/sec}$$

Since the information transmitted is maximum (as all levels are equiprobable) the source efficiency

$$\eta_s = \frac{H(S)}{H(S)_{\max}} = \frac{H(S)_{\max}}{H(S)_{\max}} = 1 = 100\%$$

EXTENSION OF ZERO-MEMORY SOURCE: Let us consider a binary

source S emitting symbols s_1 & s_2 with probabilities P_1 & P_2 such that $P_1 + P_2 = 1$

Then the 2nd extension of this binary source are is

$s_1 s_1, s_1 s_2, s_2 s_1,$ and $s_2 s_2$ & their probabilities are $P_1 P_1, P_1 P_2, P_2 P_1,$ and $P_2 P_2 \Rightarrow P_1 P_1 = P_1^2$ & $P_2 P_2 = P_2^2$

⇒ Sum of all probabilities $P_1^2 + P_1 P_2 + P_2 P_1 + P_2^2 = P_1^2 + 2P_1 P_2 + P_2^2 = 1$
 $(P_1 + P_2)^2 = P_1^2 + P_1 P_2 + P_2^2 = (1)^2 = 1^2 = 1$

⇒ Entropy of 2nd extension of source $S = \{s_1, s_2\}$ with $P = (P_1, P_2)$
 ∴ $H(S)$ = Entropy of the basic binary source

$$H(S) = \sum_{i=1}^2 P_i \log \frac{1}{P_i} = P_1 \log \frac{1}{P_1} + P_2 \log \frac{1}{P_2}$$

The entropy of 2nd extension is given by

$$H(S^2) = \sum_{i=1}^4 P_j \log \frac{1}{P_j}$$

$$P_1 P_2 = P_2 P_1$$

$$= P_1^2 \log \frac{1}{P_1^2} + P_1 P_2 \log \frac{1}{P_1 P_2} + P_2 P_1 \log \frac{1}{P_2 P_1} + P_2^2 \log \frac{1}{P_2^2}$$

$$= 2P_1^2 \log \frac{1}{P_1} + 2P_1 P_2 \log \frac{1}{P_1 P_2} + 2P_2^2 \log \frac{1}{P_2}$$

$$= 2P_1^2 \log \frac{1}{P_1} + 2P_1 P_2 \log \frac{1}{P_1 P_2} + 2P_1 P_2 \log \frac{1}{P_2} + 2P_2^2 \log \frac{1}{P_2}$$

$$= 2P_1(P_1 + P_2) \log \frac{1}{P_1} + 2P_2(P_1 + P_2) \log \frac{1}{P_2} \quad \underline{P_1 + P_2 = 1}$$

$$2(P_1 \log \frac{1}{P_1} + P_2 \log \frac{1}{P_2}) = 2H(S)$$

∴ $H(S)^n = n \cdot H(S)$

Example: A ZMS has a source alphabet - $S = \{s_1, s_2, s_3\}$ with $P = \{1/2, 1/4, 1/4\}$
 Find the entropy of this source. Also determine the entropy of its 2nd extension and verify that $H(S^2) = 2H(S)$

Soln: For the basic source $H(S) = \frac{1}{2} \log_2 \frac{1}{1/2} + 2 \times \frac{1}{4} \log_2 \frac{1}{1/4} = 1.5 \text{ bits}$

2nd extension of a source: $s_1 s_1, s_1 s_2, s_1 s_3, s_2 s_1, s_2 s_2, s_2 s_3, s_3 s_1, s_3 s_2$
 and $s_3 s_3$.
 $\frac{1}{2} \times \frac{1}{2}, \frac{1}{2} \times \frac{1}{4}, \frac{1}{2} \times \frac{1}{4}, \frac{1}{4} \times \frac{1}{2}, \frac{1}{4} \times \frac{1}{4}, \frac{1}{4} \times \frac{1}{4}, \frac{1}{4} \times \frac{1}{2}, \frac{1}{4} \times \frac{1}{4}$

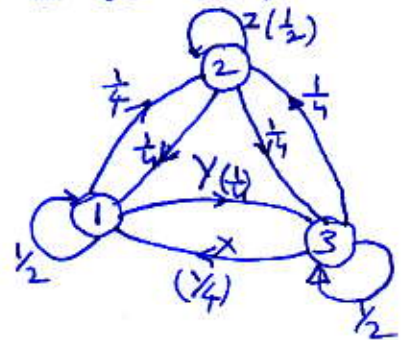
$$\therefore H(S^2) = \frac{1}{4} \log_2 4 + 4 \times \frac{1}{8} \log_2 8 + \frac{1}{16} \log_2 16 = 2 \times 1.5 \text{ bits/message} = 2 \times H(S)$$

ENTROPY OF SOURCE WITH MEMORY (MARKOV SOURCES):

In real life sources, there is intersymbol ~~interference~~ influence present such that the occurrence of x_i in zeroth position s_0 of the message depends on the previous q symbols $\{s_1, s_2, \dots, s_q\}$. Such a source is known as q th order Markov source or Markov source.

These sources are specified by a set of conditional probabilities $P(x_i | s_1, s_2, \dots, s_q)$ where x_i is the symbol in the s_0 position and each 's' has an m symbol alphabet $\{x_1, x_2, \dots, x_m\}$. Since $P(x_i)$ now depends on the earlier q symbols. The transitional probabilities may be shown in a state diagram with m^q possible states symbols.

Example: For the first order Markov source shown in figure, draw the tree diagram representing the states at the end of second symbol interval & find the corresponding probabilities. Assume $P(1) = P(2) = P(3) = 1/3$.



Soln: From the tree diagram, the symbol xx can be generated by either one of the following transitions $1 \rightarrow 1 \rightarrow 1$ or $2 \rightarrow 1 \rightarrow 1$ or $3 \rightarrow 1 \rightarrow 1$. Thus the probability of the source emitting 2 symbol seq. xx is given by

$$P(xx) = P[(1 \rightarrow 1 \rightarrow 1) \cup (2 \rightarrow 1 \rightarrow 1) \cup (3 \rightarrow 1 \rightarrow 1)]$$

Since all transition paths are disjoint

$$P(xx) = P(1 \rightarrow 1 \rightarrow 1) + P(2 \rightarrow 1 \rightarrow 1) + P(3 \rightarrow 1 \rightarrow 1) \\ = \frac{1}{3} \times \frac{1}{2} \times \frac{1}{2} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{2} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} = \frac{1}{6}$$

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$$P(XZ) = \frac{1}{3} \times \frac{1}{2} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} = \frac{1}{12}$$

$$P(XY) = \frac{1}{3} \times \frac{1}{2} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} = \frac{1}{12}$$

$$P(ZX) = \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{2} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} = \frac{1}{12}$$

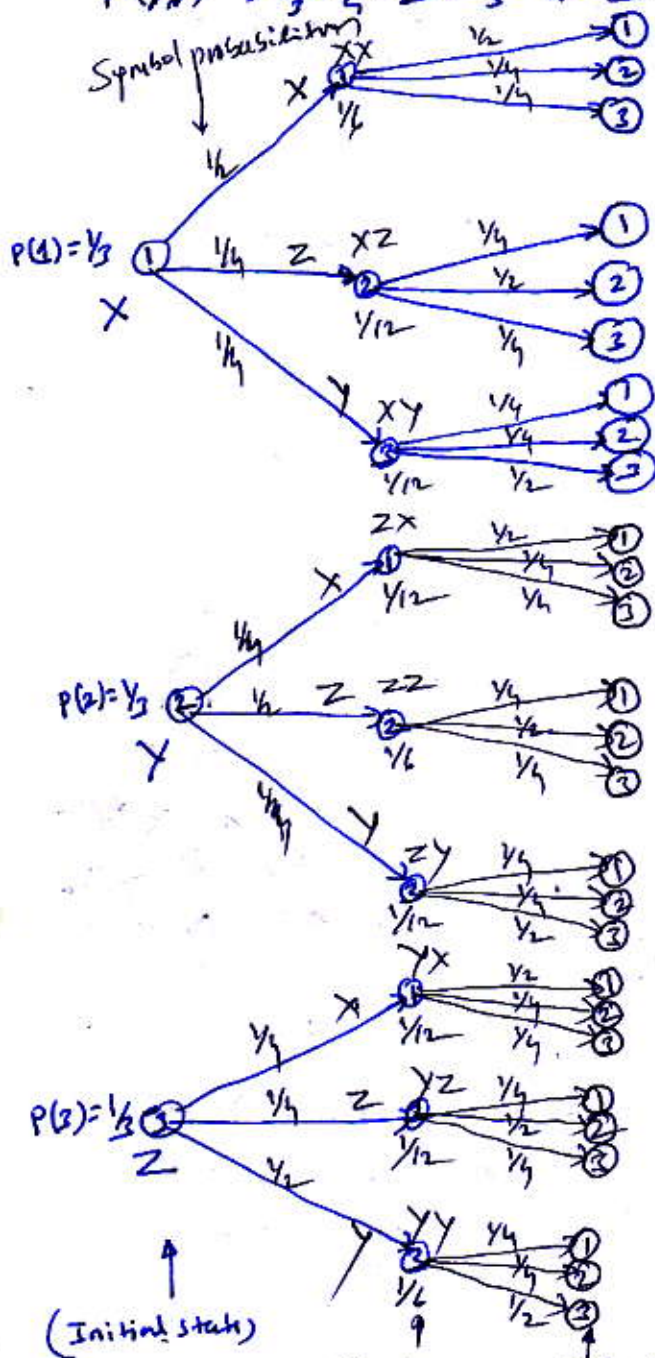
$$P(ZZ) = \frac{1}{3} \times \frac{1}{4} \times \frac{1}{2} + \frac{1}{3} \times \frac{1}{2} \times \frac{1}{2} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{2} = \frac{1}{6}$$

$$P(ZY) = \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{2} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} = \frac{1}{12}$$

$$P(YX) = \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{2} \times \frac{1}{4} = \frac{1}{12}$$

$$P(YZ) = \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{4} + \frac{1}{3} \times \frac{1}{2} \times \frac{1}{4} = \frac{1}{12}$$

$$P(YY) = \frac{1}{3} \times \frac{1}{4} \times \frac{1}{2} + \frac{1}{3} \times \frac{1}{4} \times \frac{1}{2} + \frac{1}{3} \times \frac{1}{2} \times \frac{1}{2} = \frac{1}{6}$$



Start state at the end of first symbol interval

State at the end of second symbol interval

Entropy and Information rate of Markov Sources:

The entropy of the source is defined as the weighted average of the entropy of the symbols emitted from each state. Where the entropy of state i , denoted by H_i is defined as the average info content of the symbols emitted from the i^{th} state.

$$\therefore H_i = \sum_{j=1}^n P_{ij} \log \frac{1}{P_{ij}} \text{ bits/message symbol} \quad (1)$$

The entropy of the source is then the average of the entropy of each state i.e.,

$$\begin{aligned} H &= \sum_{i=1}^n P_i H_i \\ &= \sum_{i=1}^n P_i \left[P_{ij} \log \frac{1}{P_{ij}} \right] \text{ bits/message symbol.} \end{aligned}$$

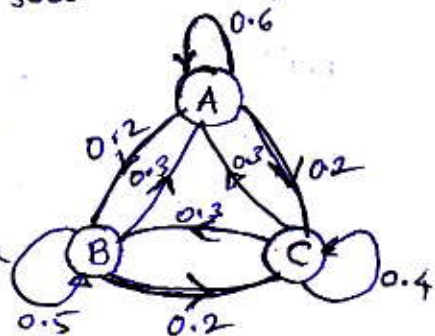
Where P_i is the probability that the source is in state i .

The average information rate R_s for the source is defined as

$$R_s = \gamma_s \cdot H \text{ bits/sec}$$

Example: For the first order Markov source shown;

- i) Find the stationary distribution
- ii) Find the entropy of each state and hence the entropy of the source
- iii) Find the entropy of the adjacent source and verify that $H(S) < H(\bar{S})$



Solution: From the state diagram, the state equations are given by

$$\text{State A: } P(A) = 0.6 P(A) + 0.3 P(B) + 0.3 P(C) \rightarrow (1)$$

$$P(B) = 0.2 P(A) + 0.5 P(B) + 0.3 P(C) \rightarrow (2)$$

$$P(C) = 0.2 P(A) + 0.2 P(B) + 0.4 P(C) \rightarrow (3)$$

$$(2) - (3) \Rightarrow P(B) - P(C) = 0.3 P(B) - 0.1 P(C).$$

$$\Rightarrow 0.7 P(B) = 0.9 P(C) \text{ or } P(B) = \frac{9}{7} P(C). \quad (4)$$

$$(1) - (2) = P(A) - P(B) = 0.4 P(A) - 0.2 P(B)$$

$$0.6 P(A) = 0.8 P(B)$$

$$\Rightarrow P(A) = \frac{4}{3} P(B)$$

$$-(5) \text{ or } P(B) = \frac{3}{4} P(A). \quad (5)$$

Equating (4) & (5) $\frac{9}{7} P(C) = \frac{3}{4} P(A)$

$$\Leftrightarrow P(C) = \frac{7 \times 3}{4 \times 9} P(A)$$

$$= \frac{7}{12} P(A) \quad \text{--- (6) or } \frac{12}{7} P(C) = P(A)$$

Also $P(A) + P(B) + P(C) = 1$ --- (7)

substituting (4), (5) & (6) in (7)

$$\cancel{P(A)} + \frac{12}{7} P(C) + \frac{9}{7} P(C) + P(C) = 1$$

$$\Rightarrow \frac{12P(C) + 9P(C) + 7P(C)}{7} = 1 \quad \text{or } 28P(C) = 7$$

$$P(C) = \frac{7}{28} = \frac{1}{4}$$

$$\therefore P(A) = \frac{12}{7} \times P(C) = \frac{12}{7} \times \frac{1}{4} = \frac{3}{7} \Rightarrow P(A) = \frac{3}{7}$$

$$\& P(B) = \frac{3}{4} P(A) = \frac{3}{4} \times \frac{3}{7} = \frac{9}{28} \Rightarrow P(B) = \frac{9}{28}$$

ii) The entropy of each state is given by eqn

$$H_i = \sum_{j=1}^n P_{ij} \log \frac{1}{P_{ij}}$$

For state A

$$H_A = \sum_{j=A}^C P_{Aj} \log \frac{1}{P_{Aj}}$$

$$= P_{AA} \log \frac{1}{P_{AA}} + P_{AB} \log \frac{1}{P_{AB}} + P_{AC} \log \frac{1}{P_{AC}}$$

$$= 0.6 \times \log \frac{1}{0.6} + 0.2 \log \frac{1}{0.2} + 0.2 \log \frac{1}{0.2}$$

$$= 1.371 \text{ bits/symbol}$$

State B

$$H_B = \sum_{j=A}^C P_{Bj} \log \frac{1}{P_{Bj}}$$

$$= P_{BA} \log \frac{1}{P_{BA}} + P_{BB} \log \frac{1}{P_{BB}} + P_{BC} \log \frac{1}{P_{BC}}$$

$$= 0.3 \log \frac{1}{0.3} + 0.5 \log \frac{1}{0.5} + 0.2 \log \frac{1}{0.2}$$

$$= 1.485$$

State C:

$$\sum_{j=A}^C P_{Cj} \log \frac{1}{P_{Cj}} = P_{CA} \log \frac{1}{P_{CA}} + P_{CB} \log \frac{1}{P_{CB}} + P_{CC} \log \frac{1}{P_{CC}}$$

$$\Rightarrow 0.3 \log \frac{1}{0.3} + 0.3 \log \frac{1}{0.3} + 0.4 \log \frac{1}{0.4} = 1.571 \text{ bits/symbol}$$

$$\therefore H(S) = H = \sum_{i=A}^C P(C_i) \cdot H(i) = P(A) H(A) + P(B) H(B) + P(C) \cdot H(C)$$

$$= \frac{3}{7} \times 1.371 + \frac{9}{28} \times 1.485 + \frac{1}{4} \times 1.571 = 1.458 \text{ bits/symbol}$$

$$(ii) \quad G_1 = H(\bar{S}) = \sum_{i=1}^3 P(m_i) \cdot \log_2 \frac{1}{P(m_i)}$$

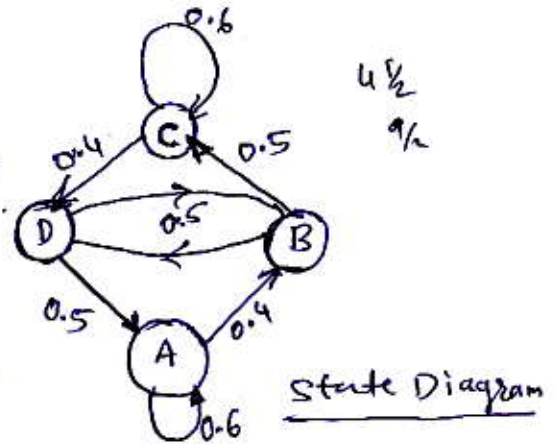
$$= P(A) \cdot \log_2 \frac{1}{P(A)} + P(B) \cdot \log_2 \frac{1}{P(B)} + P(C) \cdot \log_2 \frac{1}{P(C)}$$

$$= 1.55 \text{ bits/symbol}$$

$$\therefore H(S) < H(\bar{S})$$

Example: Consider the state diagram of the Markov source:

- i) Compute the state probabilities
- ii) Find entropy of each state
- iii) Find the entropy of the source.



Solution: State equations:

State A: $P(A) = 0.6P(A) + 0.5P(D)$ — (1)

B: $P(B) = 0.4P(A) + 0.5P(B)$ — (2)

C: $P(C) = 0.5P(B) + 0.6P(C)$ — (3)

D: $P(D) = 0.5P(D) + 0.4P(C)$ — (4)

$$0.4P(C) = 0.5P(D) \quad \text{--- (5)}$$

$$\text{or } P(C) = 1.25P(D) \quad \text{--- (6)}$$

$$= 1.25P(D) \quad \text{--- (7)}$$

Also,

From (1) & (2)

$$P(A) - P(B) = 0.2P(A)$$

$$\text{or } P(B) = 0.8P(A) \quad \text{--- (8)}$$

$$\text{or } 0.4P(A) = 0.5P(D)$$

$$\text{or } P(A) = 1.25P(D) \quad \text{--- (9)}$$

$$\therefore P(B) = P(D)$$

~~Using (4)~~ $P(D) = 0.5P(D) + 0.4 \times 1.25P(D)$

WRT: $P(A) + P(B) + P(C) + P(D) = 1$

$$1.25P(D) + P(D) + 1.25P(D) + P(D) = 1 \quad \text{or } 4.5P(D) = 1 \quad \text{or } P(D) = \frac{1}{4.5} = \frac{2}{9}$$

$$P(D) = \frac{2}{9} \therefore P(B) = \frac{2}{9}, \quad P(A) = 1.25 \times P(D) = \frac{5}{9} \times \frac{2}{9} = \frac{5}{18}$$

$$\& P(C) = \frac{5}{18}$$

iii) $H(A) = 0.971 \text{ bits/state}, \quad H(B) = 1 \text{ bit/state}, \quad H(C) = 0.971 \text{ bits/state}$
 $\& H(D) = 1 \text{ bit/state}$

$$\therefore H(S) = H = \sum_{i=1}^4 P_i H_i = P(A) \cdot H_A + P(B) \cdot H_B + P(C) \cdot H_C + P(D) \cdot H_D$$

$$\therefore H = 0.9839 \text{ bits/binary digits}$$