

Information Theory
Information Theory

25-1

Idea about: possible to achieve
 in point-to-point
 Communication Channel.

Idea about : what is possible to achieve in point to point communication channel

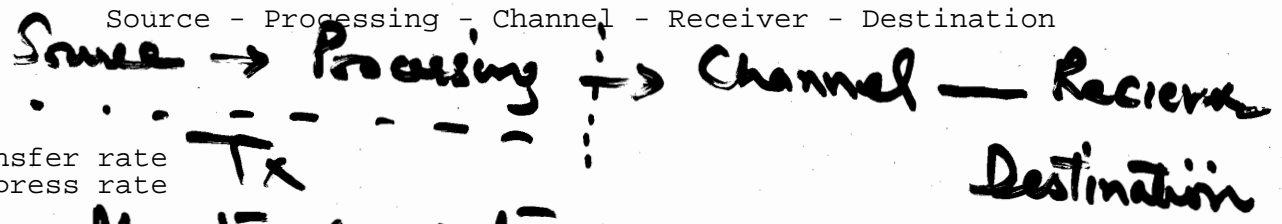
How much a signal can be compressed

How much signal can be compressed



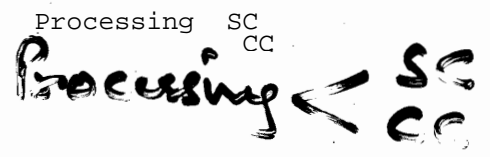
Processing : to transmit max info.
 Processing to Transmit Maximum info

Source - Processing - Channel - Receiver - Destination



Max. transfer rate
 Min. compress rate

Max Transfer rate
 Min Compress rate



SC -> to compress the source data
 CC -> transmit through Ch in efficient way (Max)

Source coding --- to compress the source data
 Channel coding --- to transmit through channel in an efficient (max) way

Source Coding

LS = 21

Example

Horse Race Car Race

8 Cars - 3 bit coding
Avg. code length 3 bit

$k_2, k_4, k_8, k_{16}, k_{64}, k_{64}, k_{64}, k_{64}$

8 cars

3 bit coding (fix)

Avg :- 3 bit

$\log_2(1/\text{prob}) = \# \text{ of bits}$

$$\log_2\left(\frac{1}{\text{prob}}\right) = \# \text{ of bits}$$

weighted avg.

Weighted Average

Most frequent events should be assigned less no. of bits

$\frac{0}{1}, \frac{10}{2}, \frac{100}{3}, \frac{1110}{4}, \frac{111100}{5}, \frac{111101}{6}, \frac{111110}{7}, \frac{111111}{8}$

Most freq. events should be assigned less no. of bits

Fewer bits \rightarrow Freq. Event

$$k_2 \times 1 + k_4 \times 2 + k_8 \times 3 + k_{16} \times 4 + \left(\frac{k}{64} \times 6\right) \times 4$$

ENTROPY

Set of events

r.v. $x \in \{1, 2, \dots, M\}$

with probs

$p(1), p(2), \dots$

$P(M)$

then the info contained

in events is

$\log(1/p(1)),$

$\log(1/p(2)), \dots$

$\log(1/p(M))$

Set of Events r.v. $x \in \{1, 2, \dots, M\}$

with probab

$p(1), p(2), \dots, p(M)$

then info in events is contained

$\log \frac{1}{p(1)}, \log \frac{1}{p(2)}, \dots, \log \frac{1}{p(M)}$

Entropy

$$H(x) = \sum_i p(i) \log \frac{1}{p(i)}$$

$$H(x) = p \log \frac{1}{p} + (1-p) \log \left(\frac{1}{1-p}\right) = -p \log p - (1-p) \log (1-p)$$

Ex: $x = \begin{cases} 1 & p \\ 0 & p-1 \end{cases}$

If $p=0$
 $H(p) = \begin{cases} 0 & p=0 \\ 1 & p=1 \end{cases}$

min # of bits required to represent the r.v. on average is $H(x)$

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Joint Entropy

Joint Entropy $H(x, y) = - \sum_x \sum_y p(x, y) \log p(x, y)$

Ex r.v. x = { a, b, c, d } with probabilities 1/2, 1/4, 1/8, 1/8

- a 1/2
b 1/4
c 1/8
d 1/8

If the code length is set equal to the H(x) i.e. the info. contained in outcomes, the avg. no. of bits in the code would be minimized

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Entropy depends on the prob. dist. of r.v. x

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1. H(x) >= 0

2. H(x) <= Lavg < H(x) + 1

Avg length of code.

Avg. length of code

Shannon Theorem Source Coding Theorem

If $H(x)$ is entropy of s.v. x then we can represent x by $H(x)$ no. of bits on average, if we apply block coding

If source generates i.i.d s.v.s x_1, x_2, \dots, x_n then total info contained would be sum of info in all x_1, \dots, x_n

$$n H(x) \leq L_M \leq n H(x) + 1$$

for M nos of symbols.

$$H(x) \leq \frac{L_M}{n} \leq H(x) + \frac{1}{n}$$

Joint Entropy

If we have two s.v.s x & y then the joint entropy would be $H(x, y) = \sum_x \sum_y p(x, y) \log \frac{1}{p(x, y)}$
 $\neq H(x) + H(y)$

Conditional Entropy

$$H(y/x) = \sum_x p(x) H(y/x=x)$$

$$H(x, y) = H(x) + H(y/x)$$

Mutual Information

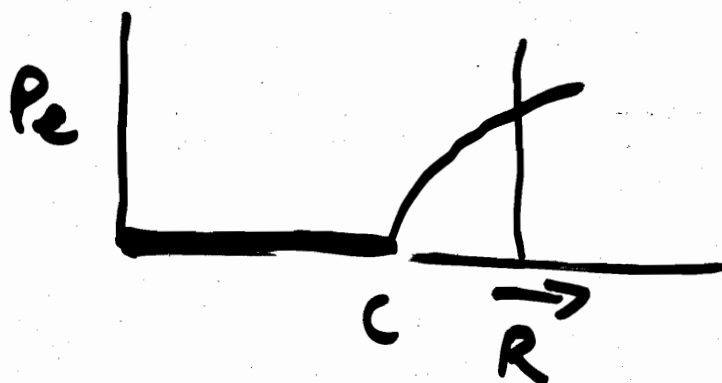
$$I(x; y) \triangleq \sum_x \sum_y p(x, y) \log \frac{p(x, y)}{p(x) p(y)}$$

Channel Coding Theorem

$$C \triangleq \max_{p(x)} I(X; Y)$$

For an $\epsilon > 0$ and $R < C$

We can transmit at rate R
with $P_e < \epsilon$.



If $R > C$
then
Reliable Communication
is not possible

P_e would be more than zero.

$\therefore P_e = 0$ if $R < C$

$P_e > 0$ if $R > C$

Examples \Rightarrow

Ex-1

Ex $X = \begin{cases} 1 & p \\ 0 & p-1 \end{cases}$ binary s.v.

$$H(X) = p \log \frac{1}{p} + (1-p) \log \left(\frac{1}{1-p} \right)$$

$$= -p \log p - (1-p) \log (1-p)$$

$$\triangleq H(p)$$

If $p = 0$

then $H(p) = \begin{cases} 0 \end{cases}$

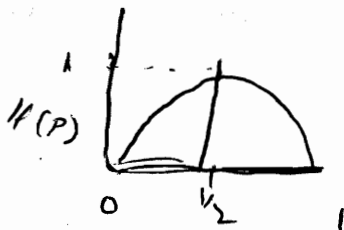
probability limit
 $\lim_{p \rightarrow 0} (1-p) \log (1-p)$
 $0 - 1 \times 0$
 0

$\frac{1}{(1-p)} \log (1-p) + \dots$
 $\frac{1}{(1-p)} \log (1-p) + \dots$

$p = 1/2$
 $\Rightarrow -\frac{1}{2} \log_2 \left(\frac{1}{2} \right) - (1-\frac{1}{2}) \log_2 \left(1-\frac{1}{2} \right)$
 $= -\frac{1}{2} (-1) - \left(\frac{1}{2} \right) (-1)$
 $= +\frac{1}{2} + \frac{1}{2} =$
 $= 1$

$\frac{1}{2} \log_2 \left(\frac{1}{2} \right) + (1-\frac{1}{2}) \log_2 \left(\frac{1}{1-\frac{1}{2}} \right)$
 $\frac{1}{2} \log_2 (2) + \frac{1}{2} \log_2 (2)$
 $= \frac{1}{2} (1) + \frac{1}{2} \times 1$
 $=$

If $p = 0$	$H(p) = 0$
$= 1$	$= 0$
$= 1/2$	$= 1$



v.v.v
Ex.

Ex-2

a s.v.x

$$\begin{cases} a & 1/2 \\ b & 1/4 \\ c & 1/8 \\ d & 1/8 \end{cases}$$

$$\begin{aligned} H(x) &= \frac{1}{2} \log_2 2 + \frac{1}{4} \log_2 4 + \frac{1}{8} \log_2 8 + \frac{1}{8} \log_2 8 \\ &= \frac{1}{2} + \frac{3}{4} + \frac{3}{8} + \frac{3}{8} \\ &= \frac{7}{4} \text{ bits} \rightarrow 1.75 \end{aligned}$$

If s.v.x \rightarrow equally probable (4 outcomes)

then $4 \times \frac{1}{4} \log_2 4$ \rightarrow uniform

$$1 \times 2 = 2 \text{ bits}$$

a	1/2	$\rightarrow \log_2 2 \rightarrow 1$	$\rightarrow 1 \text{ bit}$	0
b	1/4	$\rightarrow \log_2 4 \rightarrow 2$	$\rightarrow 2 \text{ bits}$	10
c	1/8		$\rightarrow 3 \text{ bits}$	110
d	1/8		$\rightarrow 3 \text{ bits}$	110

Average Code = $\frac{7}{4}$ bits $\rightarrow 1.75$ bits

Ex

$$X = Y \Rightarrow H(X, Y) = H(X) = H(Y)$$

$$\begin{aligned} X &\in \{0, 1\} & Y &\in \{0, 1\} \\ &p, 1-p & & q, 1-q \end{aligned}$$

Joint Entropy of X & Y

$$H(X; Y) = p q \log \frac{1}{p q} + p(1-q) \log \frac{1}{p(1-q)}$$

$$+ q(1-p) \log \left(\frac{1}{q(1-p)} \right) + (1-p)(1-q) \log \left(\frac{1}{(1-p)(1-q)} \right)$$

$$\begin{aligned} H(p) + H(q) \\ H(X) + H(Y) \end{aligned}$$

X \ Y	0	1
0	p q	p(1-q)
1	q(1-p)	(1-p)(1-q)

Ex: $x = Y$ $H(x, Y) = H(x) = H(Y)$

$X \in \{0, 1\}$ $Y \in \{0, 1\}$
 $p, p-1$ $q, 1-q$

Joint Entropy

Joint Entropy

	Y	0	1
X			
0		pq	$p(1-q)$
1		$q(1-p)$	$(1-p)(1-q)$

$H(x, Y)$

$$= pq \log \frac{1}{pq} + p(1-q) \log \frac{1}{p(1-q)} + q(1-p) \log \frac{1}{q(1-p)} + (1-p)(1-q) \log \frac{1}{(1-p)(1-q)}$$

$$= H(p) + H(q)$$

$$= H(x) + H(Y)$$

Two r.v. x & y takes 4 values each.
Joint p. d.f.

$x \backslash y$	1	2	3	4
1	$1/8$	$1/16$	$1/32$	$1/32$
2	$1/16$	$1/8$	$1/32$	$1/32$
3	$1/16$	$1/16$	$1/16$	$1/16$
4	$1/4$	0	0	0

Marginal p. d.f.

$$p(x) \rightarrow (1/8 + 1/16 + 1/16 + 1/4, 1/16 + 1/8 + 1/16 + 0, 1/32 + 1/32 + 1/16 + 0, 1/32 + 1/32 + 1/16 + 0)$$

$$\rightarrow (1/2, 1/4, 1/8, 1/8) \rightarrow \frac{1}{2} \log 2 + \frac{1}{4} \log 4 + \frac{1}{8} \log 8 + \frac{1}{8} \log 8 \Rightarrow 7/4 \text{ bits} \rightarrow H(x)$$

$$p(y) \rightarrow (1/8 + 1/16 + 1/32 + 1/32, 1/16 + 1/8 + 1/32 + 1/32, 1/16 + 1/16 + 1/16 + 1/16, 1/4 + 0 + 0 + 0)$$

$$(1/4, 1/4, 1/4, 1/4) \Rightarrow \frac{1}{4} \log 4 = 1 \times 2 = 2 \text{ bits} \rightarrow H(y)$$

$$H(x, y) = \sum_{x, y} p(x, y) \log \frac{1}{p(x, y)}$$

$$= 2 \times \frac{1}{8} \log 8 + 6 \times \frac{1}{16} \log 16 + 4 \times \frac{1}{32} \log 32 + \frac{1}{4} \log 4$$

$$H(x, y) = 27/8 \text{ bits}$$

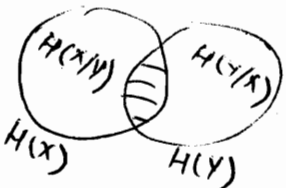
$$H(x/y) = \sum_{i=1}^4 p(Y=i) H(X|Y=i)$$

divide by 4
Joint dist.
marginal 1/4

$$= \frac{1}{4} H(4 \times (1/8, 1/16, 1/32, 1/32)) + \frac{1}{4} H(4 \times (1/16, 1/8, 1/32, 1/32)) + \frac{1}{4} H(4 \times (1/16, 1/16, 1/16, 1/16)) + \frac{1}{4} H(4 \times (1, 0, 0, 0))$$

$$\Rightarrow \frac{1}{4} \times 7.4 + \frac{1}{4} \times 7.2 + \frac{1}{4} \times 2 + \frac{1}{4} \times 0$$

$$\Rightarrow 11/8$$



$$H(x, y) = H(x) + H(y/x) \Rightarrow 27/8 - 7/4 = H(y/x) \rightarrow 13/8$$

$$H(x, y) = H(x) + H(y/x) \Rightarrow 27/8 = 7/4 + 11/8$$