



# POORNIMA

## COLLEGE OF ENGINEERING

### DETAILED LECTURE NOTES

Information measure for Continuous Random Variables:

The definition of mutual information for discrete random variable can be extended to continuous random variable.

for  $x, y \rightarrow$  random variable

$P(x, y) \rightarrow$  Probability density function

$P(x) \text{ \& } P(y) \rightarrow$  marginal pdf

Average mutual Information

It provide between two continuous random variable  $x$  and  $y$

$$I(x; y) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} P(x) P(y|x) \log \frac{P(y|x) \cdot P(x)}{P(x) \cdot P(y)} dx dy$$

for continuous variable we write  $x$  and  $y$  in the two terms.

$\rightarrow$  Avg mutual information can be carried over from discrete random variable but the concept and physical interpretation would be not.

$\rightarrow$  A continuous Random variable is actually infinite and it requires infinite no. of bit to represent.

a continuous Rv. It can be solved by differential entropy.

$$h(x) = - \int_{-\infty}^{\infty} p(x) \log p(x) dx$$
$$= \int_{-\infty}^{\infty} p(x) \log \frac{1}{p(x)} dx$$

Properties of Differential Variable:

→  $h(x+c) = h(x)$  — translation does not affect the differential entropy, where  $c$  is the constant.

→  $h(ax) = h(x) + \log |a|$

**Relative Entropy:** It measures a distance between distribution. In other words, it provides how similar are two distributions. It is called Kullback Leibler (KL) distance between two probability mass functions  $p(x)$  and  $q(x)$ .

$$D(P||Q) = \sum_{x \in X} p(x) \log \left[ \frac{p(x)}{q(x)} \right]$$



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#### Source Coding Theorem:

The conversion of the output to a Discrete memoryless Source (DMS) into a sequence of binary symbol is called source coding.

→ Application of source coding is to minimize the average bit rate required for representation of the source by reducing redundancy of the information source.

A code is a set of vectors called codeword.

#### Average Code length:

From a DMS set of alphabet  $\{x_1, x_2, \dots, x_m\}$  with corresponding probability  $\{p_1, p_2, \dots, p_m\}$  and code length  $\{l_1, l_2, \dots, l_m\}$ .

$$L = \sum_{i=1}^m p_i l_i$$

$L$  is the average code length  $L$  per source symbol is thus represented by upper term.



### Source Code Theorem:

It will help to explore efficient representation of symbols generated by a source.

#### Theorem

Suppose a DMS outputs a symbol for every  $t$  seconds. Each symbol is selected from the finite source of symbol  $i = 1, 2, \dots, L$ .

and occur with the probabilities  $p$  of  $p(x_i) = 1, 2, \dots, L$ .

The entropy of this DMS is bit per source symbol

$$H(x) = \sum_{i=1}^L p(x_i) \log_2 \frac{1}{p(x_i)} \leq \log_2 L$$

↓  
Average code word length

Example Let  $p(x_1) = 0.5$   
 $p(x_2) = 0.5$

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

$$= -0.5 \log_2 0.5 - 0.5 \log_2 0.5$$

$$= 0.5 + 0.5 = 1$$

$$\log_2 L = \log_2 2 = 1$$

$H(x) = \log_2 L$



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Example 2

$$P(x_1) = 0.5$$

$$P(x_2) = 0.25$$

$$P(x_3) = 0.25$$

$$H(x) = -0.5 \log_2 0.5 - 0.25 \log_2 0.25 - 0.25 \log_2 0.25$$

$$= -0.5 \log_2 2^{-1} - 0.25 \log_2 2^{-2} - 0.25 \log_2 2^{-2}$$

$$= 0.5 + 0.5 + 0.5 = 1.5$$

$$\text{AND } \log_2 L = \log_2 3 = \frac{\log_{10} 3}{\log_{10} 2} = \frac{0.4771}{0.301} = 1.58$$

$$\boxed{\log_2 L = 1.58}$$

### Fixed Length Code

Dms output a symbol selected from a finite set of symbol  $x_i, i = 1, 2, \dots, L$ .

The number of binary digit  $R$  required for unique coding when  $L$  is the power of 2 =

$$R = \log_2 L$$

If  $L$  is the power of 2 then  $\overset{000}{\text{---}} \overset{00}{\text{---}}$  should be considered by above equation.

$$R = \log_2 L$$

on

$$R = \lceil \log_2 L \rceil - \text{Ceiling}$$

on

If  $L$  is not power of 2  $\rightarrow \lfloor \log_2 L \rfloor + 1$  floor

if  $L$  is not the power of two

for ex  $\lceil 1.6 \rceil$  it should be round off at 2

Variable length code: When the source symbol are not equally probable, we label unequal no of bit for an efficient coding.

Example

Fixed length code Example

letter	Codeword	letter	Codeword
A	000	E	100
B	001	F	101
C	010	G	110
D	011	H	111

Variable length code Example

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

Suppose the transmitted signal A B A D C A D

fixed length code 000 001 000 011 010 000 001 Total bit - 21

Variable length code 00 010 00 100 011 00 010 Total bit - 18