	Notes
Information in Communication Systems:	
Redundancy, Source Coding and Noise	
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Redundancy	Notes
•	Notes
$lacksquare$ N symbol vocabulary has maximum possible entropy $\log_2(N)$ bits.	
Normally entropy is less than maximum entropy.	
■ ⇒ built-in Redundancy : Redundancy = 1 actual entropy	
$Redundancy = 1 - \frac{actual\ entropy}{maximum\ entropy}.$	

Redundancy Example

A source has 5 symbols, $U = \{q, w, e, r, t\}$ with probabilities:

symbol	q	w	e	r	t
probabilities	0.3	0.25	0.2	0.15	0.1

Solution

- \blacksquare Maximum Entropy= $\log_2(5)=2.32$ bits.
- Actual entropy=

 $0.3 \times 1.74 + 0.2 \times 2 + 0.2 \times 2.32 + 0.15 \times 2.74 + 0.1 \times 3.3 = 2.23$

- \blacksquare . : Redundancy=1 $-\frac{2.23}{2.32}=0.04\approx 4\%.$
- Probabilities close to uniform distribution
 - \blacksquare \Rightarrow low redundancy.

Redunda	ancy

Redundancy can be useful

■ Redundancy can help to overcome errors.

Redundancy increases amount of information sent or stored.

- Parity bit (addition of single bit)
 - Makes code 89% efficient.
 - And able to detect single bit error for every byte
- \blacksquare Hamming code (4,7)
 - $\blacksquare \ 43\%$ redundancy.
 - Corrects single error in group of 4 bits

Source or Entropy Coding

- \blacksquare Some signals might be **highly redundant** before any coding:
 - Signal can be converted into another less redundant form.
 - Process known as Source Coding.

Very old example of source coding: Morse Code

- Short codes for common letters, e.g. E is
- Long codes for uncommon letters, e.g. Z is

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Source Coding: Code Capacity	Notes
Code Capacity=Amount of information code can carry per source	
symbol	
(on average)	
$\begin{tabular}{ll} Code Capacity=\\ average number of string symbols per symbol \times how much information individual string symbol can carry \\ \end{tabular}$	
Efficiency of coding scheme:	
$\eta = rac{ ext{source entropy}}{ ext{code capacity}}$	
, ,	
Source Code Example	Notes
A source uses 5 symbols $\{q, w, e, r, t\}$ with probabilities and codes: symbol $\begin{vmatrix} q & w & e & r \\ & t & t \end{vmatrix}$	Notes
probability 0.3 0.25 0.2 0.15 0.1	
codes 1111 1110 110 10 0 Determine efficiency of the code.	
Solution Average string length =	
$0.3 \times (4 \text{ digits }) + 0.25 \times (4 \text{ digits }) + 0.2 \times (3 \text{ digits }) + 0.15 \times (2 \text{ digits })$	
$0.5 \times (4 \text{ digits }) + 0.25 \times (4 \text{ digits }) + 0.2 \times (5 \text{ digits }) + 0.16 \times (2 \text{ digits }) + 0.16 \times (2 \text{ digits })$ $+0.1 \times (1 \text{ digits }) = 3.2 \text{binary digits}.$	
Maximum entropy of a binary symbol is $\log_2(2)=1$ bit $/$ digit	
Code capacity is $3.2 \times 1 = 3.2$ bits per source symbol Entropy of source is $\sum_{k \in U} p_k h_k = 2.2228$ bits.	
: Efficiency is:	
$\frac{2.228}{3.2\times1}=69.6\% \text{ efficient}$	
Equivalent to 30.4% redundancy (not very efficient).	
Source Code Example II	N .
Modify the previous source to use shorter codes for more likely	Notes
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$	
Solution	
Average string length = $0.3\times(2 \text{ digits }) + 0.25\times(2 \text{ digits }) + 0.15\times(3 \text{ digits })$	
$0.5 \times (2 \text{ digits }) + 0.25 \times (2 \text{ digits }) + 0.15 \times (3 digits $	
Maximum entropy of a binary symbol is $\log_2(2)=1$ bit $/$ digit	
∴ Code capacity is $2.25 \times 1 = 2.25$ bits per source symbol Entropy of source is $\sum_{k \in U} p_k h_k = 2.2228$ bits.	
∴ Efficiency is:	

Equivalent to 1.0% redundancy (much more efficient than

previous codes).

Limiting Effects of Noise¹

Noise and bandwidth not considered so far...

- lacksquare Signals are usually affected by **noise** N
- \blacksquare Signal has power S and bandwidth B Hz then,

Channel Capacity upper limit on bits per second that can be sent:

$$C = B \log_2 \left(1 + \frac{S}{N} \right) \quad \text{bits/ second}.$$

Notes

Shannon-Hartley Channel Capacity

As B increases, the ${\rm SNR}{\rm =}~S/N$ can decrease.

$$C = B \log_2 \left(1 + \frac{S}{N} \right) ~~ \mathrm{bits/~second}.$$

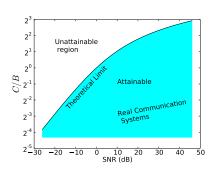
Spectral Efficiency

$$\frac{C}{B} = \log_2 \left(1 + \frac{S}{N} \right) ~~ \mathrm{bits/~second/~Hz}$$

Useful benchmark - but very difficult to realise in practice because it is a theoretical optimal value.

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Limiting Effects of Noise



$$\frac{C}{B} = \log_2(1 + {\sf SNR}) {\sf bits/second/Hz}.$$

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 $^{1}Shannon-Hartley channel capacity theorem: maximum rate of information transmission over a channel with bandwidth <math display="inline">B$ and ${\rm SNR}{=}S/N$ in bits/s.

Channel Capacity Example

Example

An audio channel has $4 \rm kHz$ bandwidth and 26dB SNR what is the upper limit of the channel capacity and the spectral efficiency?

Solution

Convert SNR = $10^{(26/10)} = 398$.

Channel capacity:

 $C = 4000 \times \log_2(1 + 398) = 34.6 \text{kbits/s}.$

Spectral efficiency is $C/4\mathrm{kHz}{=}~8~\mathrm{bits/s/Hz}.$

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Summary

- Information is communicated by communication systems.
- Smallest unit of information is a bit:
 - 1 and 0
 - YES and NO
 - ON and OFF
- Self information measures amount of information for one symbol
- Entropy measures amount of information across all symbols:
 - Entropy is the expectation of the self information for all symbols
- We also looked at maximum Entropy, efficiency, redundancy and capacity of communication systems.

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