

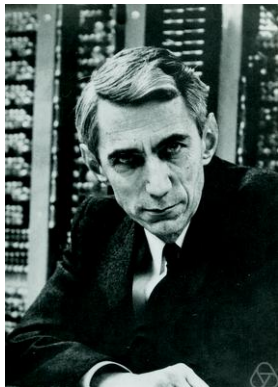
How much does this matchbox know?
(measuring information and entropy)



Alex Vlasiuk

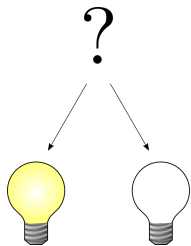
Claude Elwood Shannon

- ▶ "the father of information theory"
- ▶ ideas that have left a powerful imprint on how we think about information
- ▶ his 1948 paper is very accessible
- ▶ it is also very deep: for example, all Pierce's 1961 300-page book does is recycling Shannon's ideas



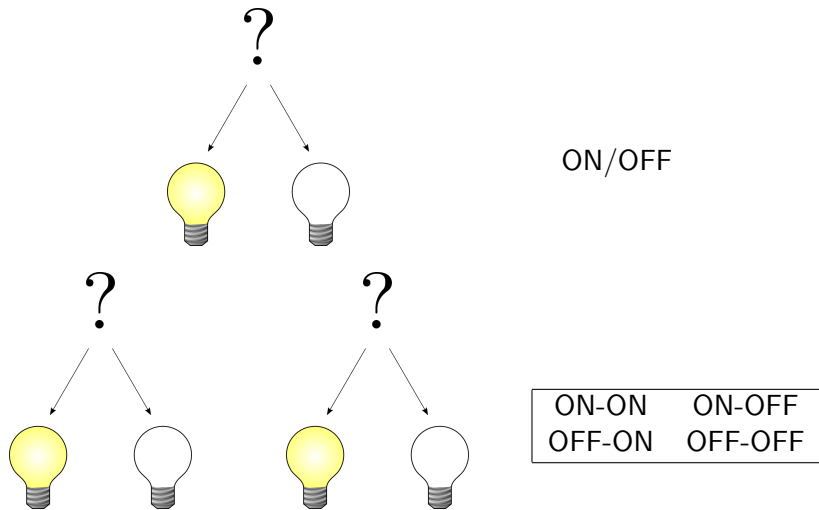
Shannon, Claude Elwood. "A mathematical theory of communication." ACM SIGMOBILE Mobile Computing and Communications Review 5.1 (2001): 3-55.

Light bulbs



ON/OFF

Light bulbs



Uncertainty of the light bulbs

For n light bulbs:

$$H = \log_2 \{\text{number of states}\} = \log_2 2^n = n.$$

These light bulbs are independent. Units of uncertainty: **bits**. The number of binary choices needed to identify the state of a system.

Causing uncertainty is opposite to having structure! Example: you can predict the next letter in the word “Floccinaucinihili

[†]estimating something as worthless

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[†]estimating something as worthless

Example: approximations to English

Using 27 (26+space) alphabet.

- ▶ symbols independent and equally probable:

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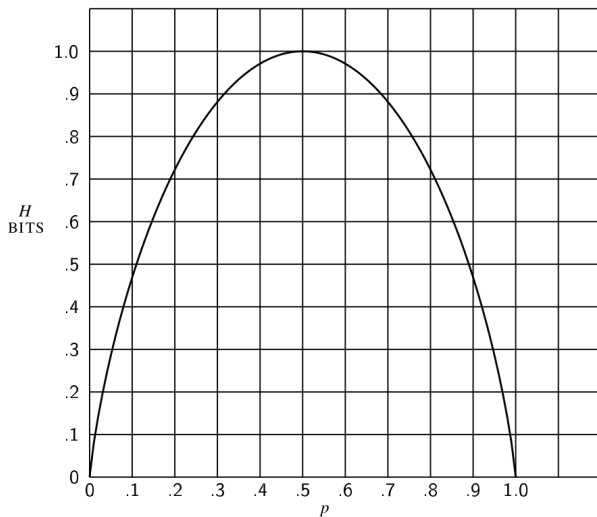
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PONDENOME OF DEMONSTURES OF THE REPTAGIN IS
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- ▶ first-order word approximation:
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- ▶ Second-order word approximation:
THE HEAD AND IN FRONTAL ATTACK ON AN ENGLISH
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UNEXPECTED

Coin toss

Probabilities of the outcomes: p and $1 - p$



Entropy

For the “coin” with probabilities p and $q = 1 - p$, the measure of uncertainty can be

$$H = -(p \log_2 p + q \log_2 q)$$

For arbitrary set of possible events with probabilities

$$p_1, p_2, \dots, p_k,$$

the measure of uncertainty in the outcome, when sampling one event:

$$H = - \sum_{i=1}^k p_i \log_2 p_i$$

A motivation

Every step selects one of k symbols in a finite “alphabet”. Suppose successive symbols are independent.

For a long message of N symbols:
around $p_1 N$ of the first, $p_2 N$ of the second, etc, so probability of this particular long message (independence!) is about

$$p = p_1^{p_1 N} p_2^{p_2 N} \dots p_k^{p_k N}$$

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$$\begin{aligned}\log_2 p &= p_1 N \log_2 p_1 + \dots + p_k N \log_2 p_k \\ &= N \sum_{i=1}^k p_i \log_2 p_i \\ &= -NH\end{aligned}$$

so

$$H = -\frac{\log_2 p}{N}$$

— average number of binary choices needed to specify one symbol in a long message.

Redundancy

{ Maximal entropy of N letters } – { Entropy of an N -letter message }

- ▶ Low redundancy (good password):

XFOML RXKHRJFFJUJ ZLPWCFWKCYJ FFJEYVKCQSGHYD
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- ▶ High redundancy (bad password):

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Redundancy \iff word puzzles!

- ▶ Redundancy = 0 \implies everything is a word! Any two-dimensional array of letters forms a crossword puzzle. If the redundancy is too high though, too many constraints for a crossword.
- ▶ Redundancy allows statistical attacks against encryption: see E. A. Poe “**The Gold-Bug**”.
- ▶ Such attacks actually only work against naive cyphers, such as the one in Gold-Bug.
- ▶ Redundancy makes it possible to read your scratched CDs.

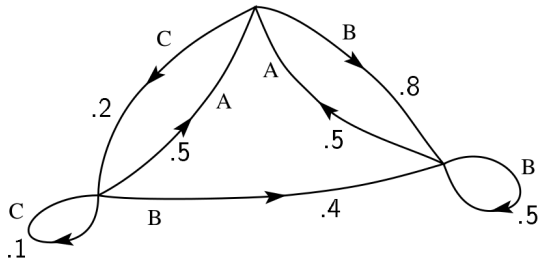
Sources with dependent symbols

Markov processes

Source is a stochastic (random) process.

Example. Alphabet: A, B, C. Transition probabilities:

$p_i(j)$		j		
	i	A	B	C
A		0	$\frac{4}{5}$	$\frac{1}{5}$
B		$\frac{1}{2}$	$\frac{1}{2}$	0
C		$\frac{1}{2}$	$\frac{2}{5}$	$\frac{1}{10}$



Entropy of a source

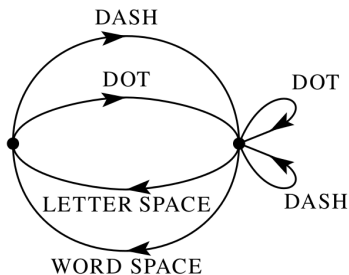
A source has states with entropies H_i and probabilities P_i , then

$$H = \sum_i P_i H_i = - \sum_{i,j} P_i p_i(j) \log_2 p_i(j)$$

As before, for N large,

$$H = - \frac{\log_2 p}{N}$$

p – probability of a typical sequence of length N



Entropies of natural languages

- ▶ another paper by Shannon: entropy of the English language is 11.82 bits per word, and since the average word has 4.5 letters, entropy is about 2.62 bits per letter
- ▶ size of char on a standard system is 8 bits
- ▶ Fabrice Bellard's compression of English:
<http://textsynth.org/sms.html>, about 15% ratio (number of output bits divided by the number of input bits)
- ▶ Kolmogorov mentions some studies of Russian with entropy around 2 bits/letter for random text with correct grammar, and 1 bit/letter for "War and piece".
- ▶ For a matchbox...

Capacity and states of a channel

Symbols: S_1, \dots, S_k with certain durations t_1, \dots, t_k . Allowed combinations of symbols are signals.

Capacity of a channel:

$$C = \frac{\log_2 N(T)}{T},$$

with T large, where $N(T)$ is the number of allowed signals of duration T .
Units: bits per second.

Noiseless channel

Theorem (the fundamental theorem for a noiseless channel)

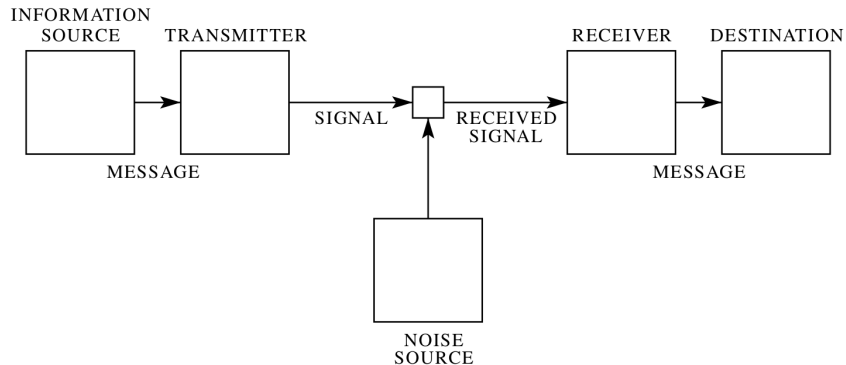
Suppose a source has entropy H bits/symbol and a channel has capacity C bits/second. Then it is possible to encode the output of the source to transmit at the average rate $\frac{C}{H} - \epsilon$ symbols/second over the channel where ϵ is arbitrarily small.

It is **not** possible to transmit at an average rate greater than $\frac{C}{H}$.

$$\frac{[C]}{[H]} = \frac{\text{bits/s}}{\text{bits/sym}} = \text{sym/s}$$

The proof involves constructing an explicit code that achieves the required rate: Shannon-Fano coding.

Overall setting



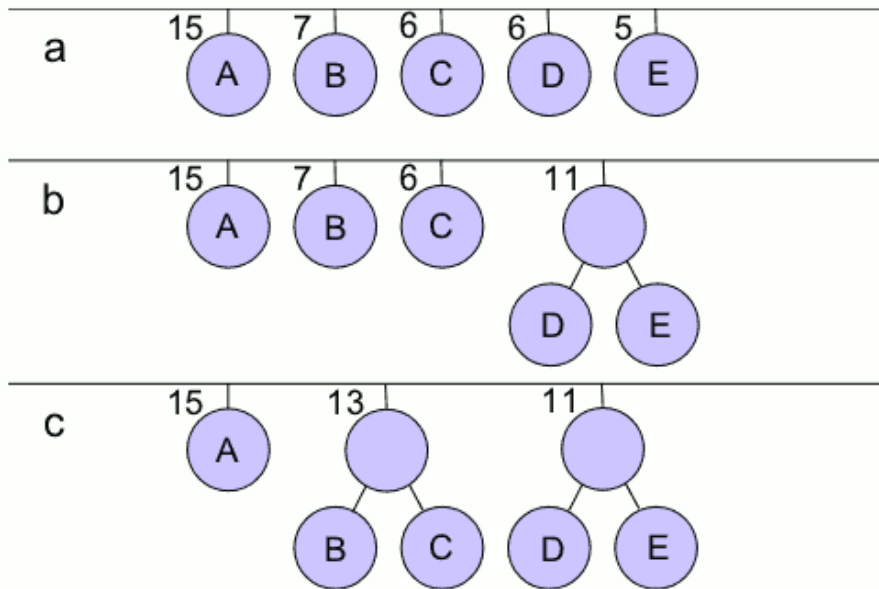
Huffman coding

Algorithm

1. Create a leaf node for each symbol and add it to a **priority queue**, by decreasing frequency.
2. While there is more than one node in the queue:
 - ▶ Remove the two nodes of lowest probability or frequency from the queue
 - ▶ Prepend 0 and 1 respectively to any code already assigned to these nodes
 - ▶ Create a new internal node with these two nodes as children and with probability equal to the sum of the two nodes' probabilities.
 - ▶ Add the new node to the queue.

The remaining node is the root node and the tree is complete.

Huffman coding example

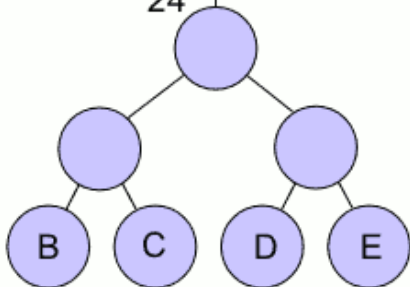


d

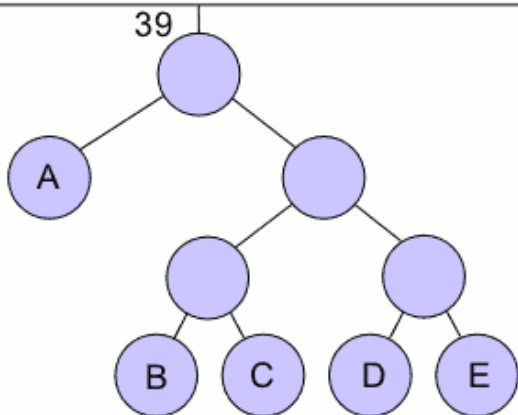
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24



e



Codewords:

A	0		C	101		E	111
B	100		D	110			

$$\frac{1 \text{ bit} \cdot 15 + 3 \text{ bits} \cdot (7 + 6 + 6 + 5)}{39 \text{ symbols}} \approx 2.23 \text{ bits per letter.}$$

Conclusions

- ▶ Entropy $H = -\sum_i p_i \log_2 p_i$ is a measure of uncertainty/information
- ▶ Allows to determine optimal compression
- ▶ Can be used to introduce redundancy, increasing reliability of the encoding

Thanks!

