

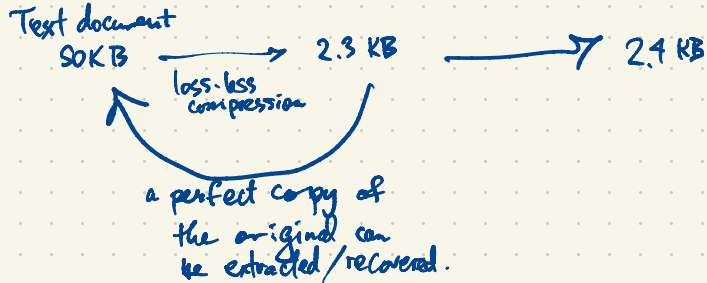
A 3D perspective view of a grid of rectangular blocks. Most blocks are grey, but one block in the upper-left quadrant is gold. The blocks are arranged in a staggered pattern, creating a sense of depth and perspective. The lighting is soft, casting gentle shadows between the blocks.

Information Theory

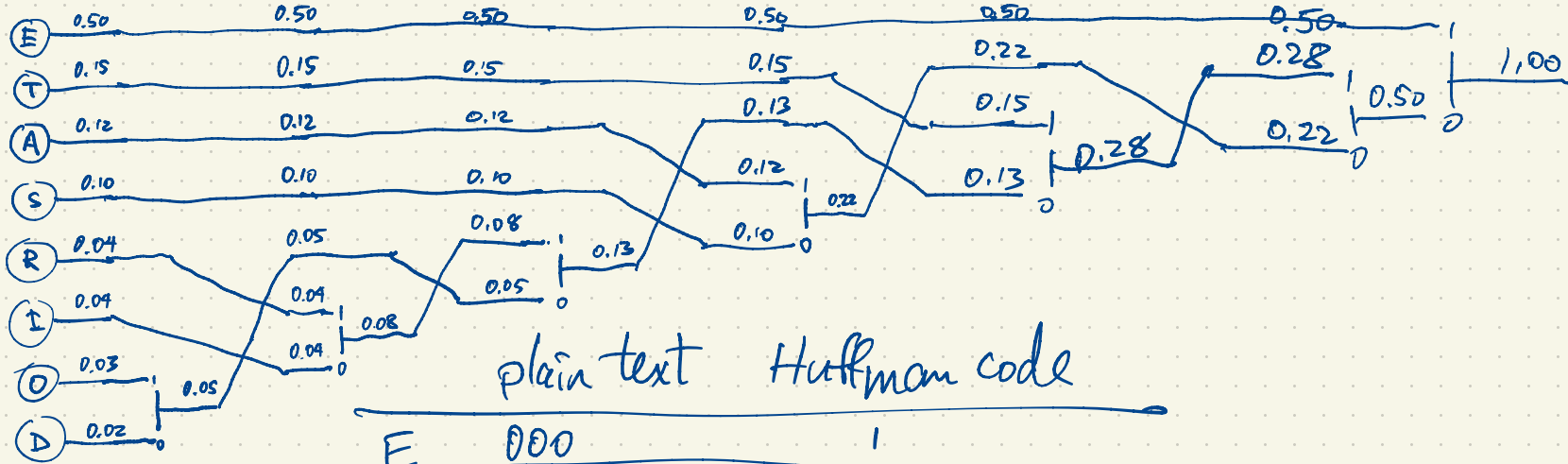
Book I

Information theory:

- Shannon information (statistical measurement of information content; classical information theory)
- Kolmogorov information (algorithmic information)
- Quantum information



Consider an information stream composed of E, T, A, S, R, I, O, D
(stream of independent letters) freq. 0.50, 0.15, ..., 0.02



plain text	Huffman code
E	000
T	001
A	010
S	011
R	100
I	101
O	110
D	111

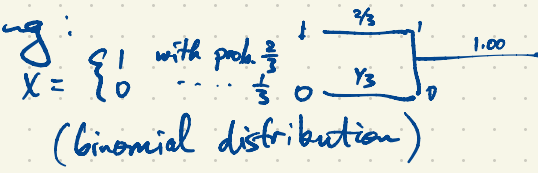
Huffman encoding:
 Encode STEER
 Decoding

as 000111101011
 S T E E R

A string of n characters is represented as $3n$ bits (plain text) which the Huffman code compresses to $2.26n$ bits. (75.37% of original)

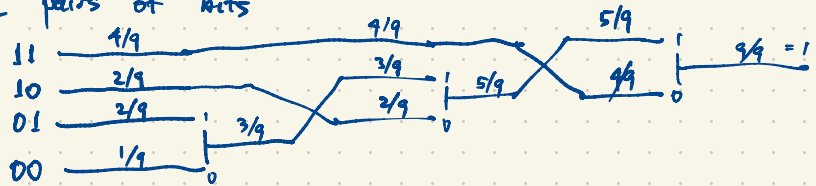
Example 2 of Huffman coding:
Stream of 0's and 1's

$\frac{1}{3}$ $\frac{2}{3}$



Plain text Huffman code
1 1
0 0
No compression

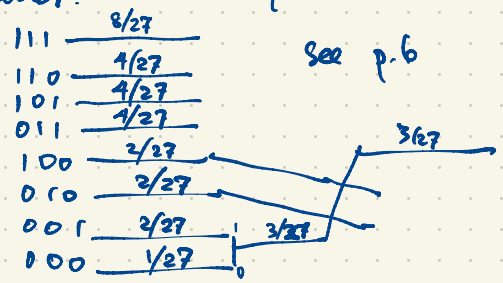
Take pairs of bits



Plain text	Huffman code
11	0
10	10
01	111
00	110

On average, a plain text file of n bits encodes as $\frac{17}{18}n \approx 0.9444n$ bits.

Better: encode triples of bits



Plain text	Huffman code
111	11
110	00
101	101
011	0111
100	100
010	0110
001	0101
000	0100

On average, n bits is encoded as $\frac{76}{27}n$ bits
 $\approx 0.9383n$ bits.

What is the limit of the compression ratio (as the block size $\rightarrow \infty$)?
 $0.9183n$ bits is the limit for compressing n bits from this stream

Shannon's first theorem showed that this stream has an entropy of

$$H(X) = \frac{1}{3} \log \frac{1}{\frac{1}{3}} + \frac{2}{3} \log \frac{1}{\frac{2}{3}} \approx 0.9183$$

Example 1 Huffman code with blocksize 1 character gives n bits $\rightarrow \frac{2.26}{3} n$ bits $\approx 0.753 n$ bits

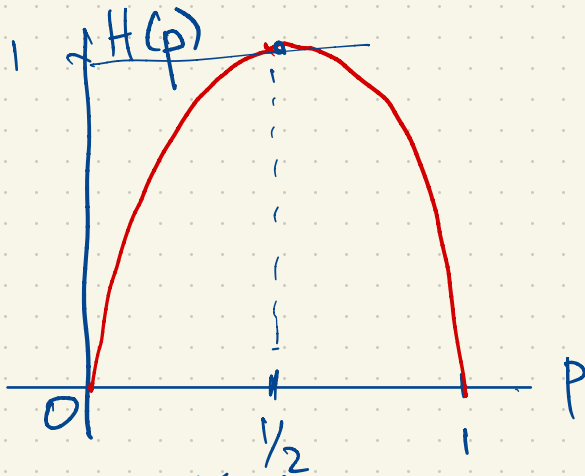
Entropy: $\sum_i p_i \log_2 \frac{1}{p_i} \approx 1.55678$ bits per character

$p_i = 0.5, 0.15, \dots, 0.12$ ($i=1, 2, \dots, 8$)

Compare: plain text encoding of characters requires 3 bits.

Binary entropy function: A biased coin has heads with prob. p and tails with prob. $1-p$, with independent tosses. $0 < p < 1$

$H(\text{coin}) = p \log_2 \frac{1}{p} + (1-p) \log_2 \frac{1}{1-p}$ = no. of bits (on average) to express the outcome of each coin flip.



Recall: If X is a random variable with outcomes $X = x_i$ ($1 \leq i \leq n$) with prob. p_i ($\sum p_i = 1$), then the binary entropy of X is $H_2(X) = \sum_{i=1}^n p_i \log_2 \frac{1}{p_i}$ = no. of bits on average required to express observed values of X .

When expressing information in base q , the q -ary entropy function $H_q(X) = \sum_{i=1}^n p_i \log_q \left(\frac{1}{p_i} \right) = \frac{1}{\log_2 q} H_2(X)$

Starting Friday, move to CR144

Eq. A byte is 8 bits. $2^8 = 256$

If X can be encoded using N bits then it takes $\frac{N}{8}$ bytes.

If I buy a deck of cards its entropy is 0 in the sense that no information is required to express the order of the deck. After shuffling the deck, it takes 225.58 bits to express the order.

↑ ignore jokers. $\log_2 52! \approx 225.58$
(about 68 decimals).

2nd Law of Thermodynamics

Watch the 7-8 minute video linked on course website

p. 49 Shannon's Source Coding Theorem (for channel without noise)

A channel is used to send a stream of symbols e.g. 0's and 1's reliably at a certain number of bits per second. Information coming from a source X has finitely many outcomes with entropy $H(X) = H_{\text{avg}}(X)$ bits per symbol e.g. x_1, \dots, x_n or A, B, C, D, ...

This information can be reliably sent and received at a maximum rate $\frac{C}{H}$ $\left. \begin{array}{l} \text{bits/sec} \\ \text{bits/symbol} \end{array} \right\}$ symbols/sec.

Eq. X is a stream of characters E, T, ..., D (first example) with prob. 0.50, 0.15, ..., 0.02, $H(X) = 1.55$ bits/char.

If I transmit info. from this source using a ^{noiseless} channel with capacity 31 bits/sec. then

I can safely transmit less than $\frac{C}{H} = \frac{31 \text{ bits/sec}}{1.55 \text{ bits/char}} = 20$ char./sec.

We can get within any pos. ϵ of this optimal rate i.e. $20 - \epsilon$.