Information Theory Chapter3: Source Coding

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Outline Chapter 2: Source Coding

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Communication Channel

from an information theoretic point of view





Variable Length Encoding

Given some source alphabet $\mathcal{X} = \{x_1, \dots, x_m\}$, code alphabet $\mathcal{Y} = \{y_1, \dots, y_d\}$.

Aim:

For each character x_1, \ldots, x_m find a code word formed over \mathcal{Y} .

Formally: Map each character $x_i \in \mathcal{X}$ uniquely onto a "word" over \mathcal{Y} .

Definition 3.1.

An injective mapping

$$g: \mathcal{X} \rightarrow \bigcup_{\ell=0}^{\infty} \mathcal{Y}^{\ell}: x_i \mapsto g(x_i) = (w_{i1}, \dots, w_{in_i})$$

is called *encoding*. $g(x_i) = (w_{i1}, \ldots, w_{in_i})$ is called *code word* of character x_i , n_i is called *length* of code word *i*.



Variable Length Encoding

Example:

	g_1	g ₂	g 3	g 4
а	1	1	0	0
b	0	10	10	01
с	1	100	110	10
d	00	1000	111	11
	no encoding	encoding,	encoding,	encoding,
		words are separable	shorter,	even shorter,
			words separable	not separable

Hence, separability of concatenated words over ${\mathcal Y}$ is important.



Variable Length Encoding

Definition 3.2.

An encoding g is called *uniquely decodable (u.d.)* or *uniquely decipherable*, if the mapping

$$G: igcup_{\ell=0}^{\infty} \mathcal{X}^\ell
ightarrow igcup_{\ell=0}^{\infty} \mathcal{Y}^\ell: ig(a_1,\ldots,a_k) \mapsto (g(a_1),\ldots,g(a_k)ig)$$

is injectiv.

Example:

Use the previous encoding g_3

	g ₃	111100011011100010
а	0	1 1 1 1 0 0 0 1 1 0 1 1 1 0 0 0 1 0
b	10	1 1 1 1 0 0 0 1 1 0 1 1 1 0 0 0 1 0
с	110	1 1 1 1 0 0 0 1 1 0 1 1 1 0 0 0 1 0
d	111	dbaacdaaab

 $(g_3 \text{ is a so called prefix code})$



Prefix Codes

Definition 3.3.

A code is called *prefix code*, if no complete code word is prefix of some other code word, i.e., no code word evolves from continuing some other.

Formally: $a \in \mathcal{Y}^k$ is called prefix of $b \in \mathcal{Y}^l$, $k \leq l$, if there is some $c \in \mathcal{Y}^{l-k}$ such that b = (a, c).

Theorem 3.4.

Prefix codes are uniquely decodable.

More properties:

- Prefix codes are easy to construct based on the code word lengths.
- Decoding of prefix codes is fast and requires no memory storage.

Next aim: characterize uniquely decodable codes by their code word lengths.



Kraft-McMillan Theorem

Theorem 3.5. (a) McMillan (1959), b) Kraft (1949))

a) All uniquely decodable codes with code word lengths n_1, \ldots, n_m satisfy



b) Conversely, if $n_1, \ldots, n_m \in \mathbb{N}$ are such that $\sum_{j=1}^m d^{-n_j} \leq 1$, then there exists a u.d. code (even a prefix code) with code word lengths n_1, \ldots, n_m .

Example:

	g 3	g_4
а	0	0
b	10	01
с	110	10
d	111	11
	u.d.	not u.d.

For g_3 : $2^{-1} + 2^{-2} + 2^{-3} + 2^{-3} = 1$ For g_4 : $2^{-1} + 2^{-2} + 2^{-2} + 2^{-2} = 5/4 > 1$

 g_4 is not u.d., there is no u.d. code with code word lengths 1,2,2,2.



Kraft-McMillan Theorem, Proof of b)

Assume $n_1 = n_2 = 2$, $n_3 = n_4 = n_5 = 3$, $n_6 = 4$. Then $\sum i = 1^6 = 15/16 < 1$

Construct a prefix code by a binary code tree as follows.



The corresponding code is given as



Average Code Word Length

Given a code $g(x_1), \ldots, g(x_m)$ with code word lengths n_1, \ldots, n_m . Question: What is a reasonable measure of the "length of a code"?

Definition 3.6.

The *expected code word length* is defined as

$$\bar{n} = \bar{n}(g) = \sum_{j=1}^{m} n_j p_j = \sum_{j=1}^{m} n_j P(X = x_j)$$

Example:

	pi	g 2	g 3
а	1/2	1	0
b	1/4	10	10
С	1/8	100	110
d	1/8	1000	111
$\bar{n}(g)$		15/8	14/8
H(X)	14/8		



Noiseless Coding Theorem, Shannon (1949)

Theorem 3.7.

Let random variable X describe a source with distribution $P(X = x_i) = p_i, i = 1, ..., m$. Let the code alphabet $\mathcal{Y} = \{y_1, ..., y_d\}$ have size d.

a) Each u.d. code g with code word lengths n_1, \ldots, n_m satisfies

 $\bar{n}(g) \geq H(X)/\log d.$

b) Conversely, there is a prefix code, hence a u.d. code g with

 $\bar{n}(g) \leq H(X)/\log d + 1.$



Proof of a)

For any u.d. code it holds by McMillan's Theorem that

$$\begin{aligned} \frac{\mathcal{H}(X)}{\log d} - \bar{n}(g) &= \frac{1}{\log d} \sum_{j=1}^{m} p_j \log \frac{1}{p_j} - \sum_{j=1}^{m} p_j n_j \\ &= \frac{1}{\log d} \sum_{j=1}^{m} p_j \log \frac{1}{p_j} + \sum_{j=1}^{m} p_j \frac{\log d^{-n_j}}{\log d} \\ &= \frac{1}{\log d} \sum_{j=1}^{m} p_j \log \frac{d^{-n_j}}{p_j} \\ &= \frac{\log e}{\log d} \sum_{j=1}^{m} p_j \ln \frac{d^{-n_j}}{p_j} \\ &\leq \frac{\log e}{\log d} \sum_{j=1}^{m} p_j \left(\frac{d^{-n_j}}{p_j} - 1 \right) \\ &\leq \frac{\log e}{\log d} \sum_{j=1}^{m} \left(d^{-n_j} - p_j \right) \leq 0 \end{aligned}$$



Proof of b) Shannon-Fano Coding

W.l.o.g. assume that $p_j > 0$ for all j.

Choose integers n_j such that $d^{-n_j} \leq p_j < d^{-n_j+1}$ for all j. Then

$$\sum_{j=1}^m d^{-n_j} \leq \sum_{j=1}^m p_j \leq 1$$

such that by Kraft's Theorem a u.d. code g exists. Furthermore,

$$\log p_j < (-n_j+1)\log d$$

holds by construction. Hence

$$\sum_{j=1}^m p_j \log p_j < (\log d) \sum_{j=1}^m p_j (-n_j + 1),$$

equivalently,

$$H(X) > (\log d) (\overline{n}(g) - 1).$$



Compact Codes

Is there always a u.d. code g with

$$\bar{n}(g) = H(X)/\log d?$$

No! Check the previous proof. Equality holds if and only if $p_j = 2^{-n_j}$ for all j = 1, ..., m.

Example. Consider binary codes, i.e., d = 2. $\mathcal{X} = \{a, b\}$, $p_1 = 0.6$, $p_2 = 0.4$. The shortest possible code is g(a) = (0), g(b) = (1).

$$H(X) = -0.6 \log_2 0.6 - 0.4 \log_2 0.4 = 0.97095$$

 $\bar{n}(g) = 1.$

Definition 3.8.

Any code of shortest possible average code word length is called *compact*.

How to construct compact codes?



Huffman Coding





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Huffman Coding



A compact code g^* is given by:

Character:	а	b	с	d	е	f	g	h
Code word:	01111	01110	0110	111	110	010	10	00

It holds (log to the base 2):

 $\bar{n}(g^*) = 5 \cdot 0.05 + \dots + 2 \cdot 0.3 = 2.75$ $H(X) = -0.05 \cdot \log_2 0.05 - \dots - 0.3 \cdot \log_2 0.3 = 2.7087$



Block Codes for Stationary Sources

Encode blocks/words of length N by words over the code alphabet \mathcal{Y} . Assume that blocks are generated by a stationary source, a stationary sequence of random variables $\{X_n\}_{n\in\mathbb{N}}$. Notation for a block code:

$$g^{(N)}:\mathcal{X}^N o igcup_{\ell=0}^\infty\mathcal{Y}^\ell$$

Block codes are "normal" variabel length codes over the extended alphabet \mathcal{X}^N .

A fair measure of the "length" of a block code is the average code word length per character

$$\bar{n}(g^{(N)})/N.$$

The lower Shannon bound, namely the entropy of the source, is asymptotically $(N \to \infty)$ attained by suitable block codes, as is shown in the following.



Noiseless Coding Theorem for Block Codes

Theorem 3.9.

Let $X = \{X_n\}_{n \in \mathbb{N}}$ be a stationary source. Let the code alphabet $\mathcal{Y} = \{y_1, \dots, y_d\}$ have size d.

a) Each u.d. block code $g^{(N)}$ satisfies

$$\frac{\bar{n}(g^{(N)})}{N} \geq \frac{H(X_1, \dots, X_N)}{N \log d}$$

b) Conversely, there is a prefix block code, hence a u.d. block code $g^{(N)}$ with

$$\frac{\bar{n}(g^{(N)})}{N} \leq \frac{H(X_1,\ldots,X_N)}{N\,\log d} + \frac{1}{N}.$$

Hence, in the limit as $N \to \infty$: There is a sequence of u.d. block codes $g^{(N)}$ such that

$$\lim_{N\to\infty}\frac{\bar{n}(g^{(N)})}{N}=\frac{H_{\infty}(X)}{\log d}$$



Huffman Block Coding

In principle, Huffman encoding can be applied to block codes. However, problems include

- The size of the Huffman table is m^N, thus growing exponentially with the block length.
- ▶ The code table needs to be transmitted to the receiver.
- The source statistics are assumed to be stationary. No adaptivity to to changing probabilities.
- Encoding and decoding only per block. Delays occur at the beginning and end. Padding may be necessary.

"Arithmetic coding" avoids these shortcomings.



Assume that

- Message (x_{i1},...,x_{iN}), x_{ij} ∈ X, j = 1,..., N is generated by some source {X_n}_{n∈ℕ}.
- All (conditional) probabilities

$$P(X_n = x_{i_n} \mid X_1 = x_{i_1}, \dots, X_{n-1} = x_{i_{n-1}}) = p(i_n \mid i_1, \dots, i_{n-1}),$$

 $x_{i_1}, \ldots, x_{i_n} \in \mathcal{X}$, $n = 1, \ldots, N$, are known to the encoder and decoder, or can be estimated.

Then,

$$P(X_1 = x_{i_1}, \ldots, X_n = x_{i_n}) = p(i_1, \ldots, i_n)$$

can be easily computed as

$$p(i_1,...,i_n) = p(i_n \mid i_1,...,i_{n-1}) \cdot p(i_1,...,i_{n-1})$$



Iteratively construct intervals

Initialization, n = 1: (c(1) = 0, c(m + 1) = 1)

$$I(j) = [c(j), c(j+1)), \quad c(j) = \sum_{i=1}^{j-1} p(i), \ j = 1, \dots, m$$

(cumulative probabilities)

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Recursion over $n = 2, \ldots, N$:

$$I(i_{1},...,i_{n}) = \left[c(i_{1},...,i_{n-1}) + \sum_{i=1}^{i_{n}-1} p(i_{n} \mid i_{1},...,i_{n-1}) \cdot p(i_{1},...,i_{n-1})\right) \\ c(i_{1},...,i_{n-1}) + \sum_{i=1}^{i_{n}} p(i_{n} \mid i_{1},...,i_{n-1}) \cdot p(i_{1},...,i_{n-1})\right)$$

Program code available from Togneri, deSilva, p. 151, 152



Example.





Encode message $(x_{i_1}, \ldots, x_{i_N})$ by the binary representation of some binary number in the interval $I(i_1, \ldots, i_n)$.

A scheme which usually works quite well is as follows.

Let $l = l(i_1, \ldots, i_n)$ and $r = r(i_1, \ldots, i_n)$ denote the left and right bound of the corresponding interval. Carry out the binary expansion of l and runtil until they differ. Since l < r, at the first place they differ there will be a 0 in the expansion of l and a 1 in the expansion of r. The number $0.a_1a_2...a_{t-1}1$ falls within the interval and requires the least number of bits.

$$(a_1a_2\ldots a_{t-1}1)$$
 is the encoding of (x_{i_1},\ldots,x_{i_N}) .

The probability of occurrence of message $(x_{i_1}, \ldots, x_{i_N})$ is equal to the length of the representing interval. Approximately

$$-\log_2 p(i_1,\ldots,i_n)$$

bits are needed to represent the interval, which is close to optimal.



Example. Assume a memoryless source with 4 characters and probabilities

$$\begin{array}{c|ccccc} x_i & a & b & c & d \\ \hline P(X_n = x_i) & 0.3 & 0.4 & 0.1 & 0.2 \end{array}$$

Encode the word (bad):

а	b	с	d	
0.3	0.4	Q.1	0.2	
- ba	bb	bc	bd	
0.12	0.16	0.04	0.08	
baa	bab	bac	bād -	
0.036	0.048	0.012	0.024	
		0.3	96 0.42	2

$$(bad) = [0.396, 0.42)$$

 $0.396 = 0.01100...$ $0.420 = 0.01101...$
 $(bad) = (01101)$

