

Information Theory

Data compression

CoCoNut, 2016
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	Compression “SOURCE CODING”	Error correction “CHANNEL CODING”
Information theory	Source coding theorem Kraft-McMillan ineq.	Channel coding theorem Channel capacity
Coding methods	Symbol codes Huffman codes	Hamming codes Reed Solomon codes LDPC codes

- The **Shannon information content** of an outcome x_i :

$$h(x_i) = \log_2 \frac{1}{p(x_i)}$$

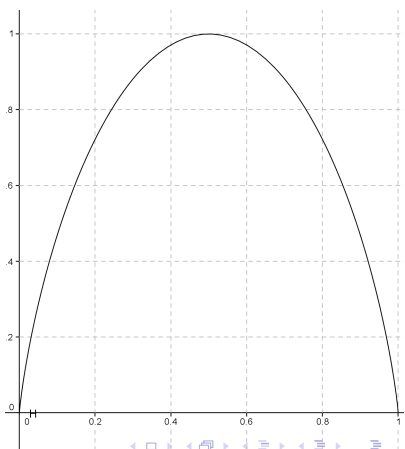
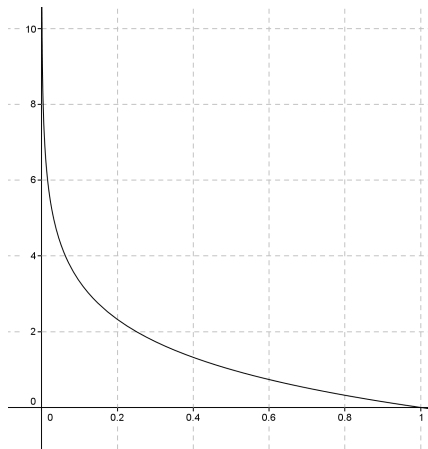
It is measured in bits.

- The **entropy** $H(X)$ is a sensible measure of the average amount of information contained in each outcome we obtain:

$$H(X) = \sum_{i=1}^m p(x_i) \log_2 \frac{1}{p(x_i)}$$

If X is a r.v. such that $\Pr(X = 0) = p$ and $\Pr(X = 1) = 1 - p$, then the entropy of X is

$$H(p) = -p \log_2 p - (1 - p) \log_2 (1 - p) \quad \text{binary entropy function}$$



$$I(X; Y) = H(X) - H(X|Y) = H(Y) - H(Y|X)$$

The **capacity** of a DMC is

$$C = \max_{p(x)} I(X; Y).$$

Theorem (Channel coding theorem (Informal))

The maximum rate R of information over a channel with arbitrarily low error probability is given by its channel capacity C .

- The capacity measures the rate at which block of data can be communicated over the channel with arbitrarily small probability of error.

- ★ **Lossless** codes which achieve **compression** (and decompression) without errors
- ★ **Variable-length** codes which encode one symbol at time
- ◇ **Theory**: How well these source codes perform?
- ◇ **Practice**: How can we construct a good source code?

- \mathcal{A}^N the set of all strings of length N over an alphabet \mathcal{A}
- \mathcal{A}^+ the set of strings of all finite length over \mathcal{A}

DEFINITION: Let X be an m -ary source with $\mathcal{A} = \{a_1, \dots, a_m\}$ and $p(a_1), \dots, p(a_m)$. A **binary source code** C for X is an encoding map:

$$\begin{aligned}\mathcal{A} &\longrightarrow \{0, 1\}^+ \\ a_i &\longmapsto c(a_i).\end{aligned}$$

The **extended code** C^+ is a mapping:

$$\begin{aligned}\mathcal{A}^N &\longrightarrow \{0, 1\}^+ \\ \mathbf{a} = (a_1 a_2 \dots a_N) &\longmapsto c(a_1)c(a_2) \dots c(a_N) = \mathbf{c}^+(\mathbf{a}),\end{aligned}$$

obtained from C by concatenation.

- ★ Let $c(\mathbf{a})$, we denote its length by $l(\mathbf{a})$ and $l_i = l(a_i)$, for $a_i \in \mathcal{A}$.

- ★ **INTUITION**: to achieve compression, on average, we assign shorter encodings to the more probable outcomes and longer encodings to the less probable.
- ★ Let X be an m -ary source and $\{p_1, \dots, p_m\}$ its probability distribution. The **expected length** of a code C for X is

$$L(C, X) = \sum_{i=1}^m p_i \cdot l_i,$$

where $\{l_1, \dots, l_m\}$ is the set of codewords lengths.

The three properties that a source code has to achieve are:

- 1 **Correct decoding**: any encoded string must have a unique decoding
- 2 **Easy decoding** : the decoding procedure has to be easy/efficient
- 3 **Small expected length**: the code should achieve as much compression as possible

Example

Given three different symbols a, b, c , such that

$$\Pr(a) = 0.5 \quad \Pr(b) = 0.25 \quad \Pr(c) = 0.25.$$

With encoding:

$$\begin{aligned} a &\mapsto 0 \\ b &\mapsto 01 \\ c &\mapsto 10 \end{aligned}$$

- ★ A code C is said to be **uniquely decodable** if $\forall \mathbf{x}, \mathbf{y} \in \mathcal{A}^+$, such that $\mathbf{x} \neq \mathbf{y}$, we have $c^+(\mathbf{x}) \neq c^+(\mathbf{y})$.
 - ◇ A source code is said to be **nonsingular** if every element of \mathcal{A} is mapped into a different string of $\{0, 1\}^+$, i.e.

$$\text{if } a_i \neq a_j \text{ then } c(a_i) \neq c(a_j).$$

Example

Consider the u.d. code $C = \{1, 10, 000, 100\}$, encoding the alphabet symbols a, b, c, d respectively. Suppose to receive

1

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 - ◇ Note that an instantaneous code is also uniquely decodable, but not the other way around.

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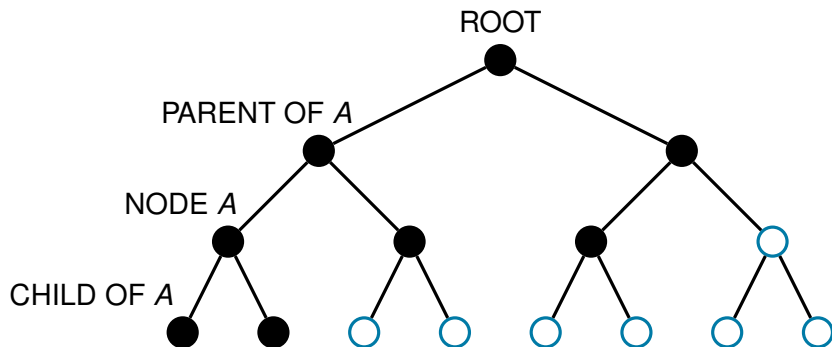
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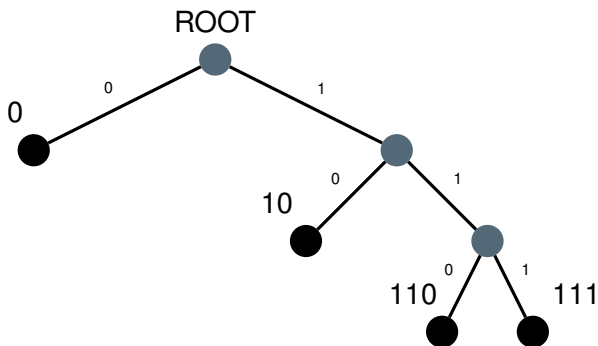
A code is instantaneous if and only if it is a prefix code.



- ♣ For every binary prefix code, there exists at least one binary tree such that each codeword corresponds to the sequence of labels of an unique path from the root to a leaf.
- ♣ Conversely, every binary tree defines a prefix code. The codewords of this prefix code are defined as the sequences of labels of each path from the root to each leaf of the coding tree.

Let C be the code $\{0, 10, 110, 111\}$

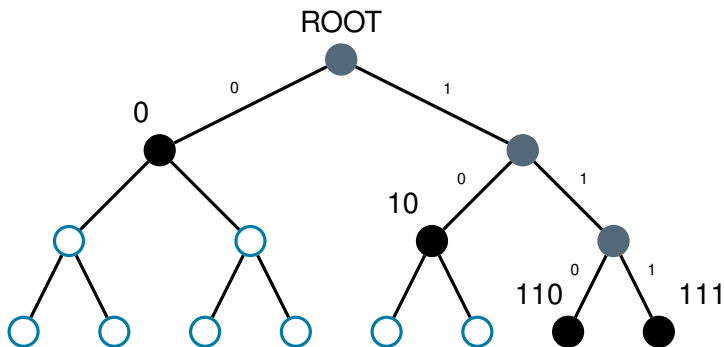
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Note that $l_i = \log_2 \frac{1}{p_i}$ ($p_i = 2^{-l_i}$)

Kraft-McMillan

- ① For each uniquely decodable binary code $C = \{\mathbf{c}_1, \dots, \mathbf{c}_m\}$, the codeword lengths must satisfy

$$\sum_{i=1}^m 2^{-l_i} \leq 1.$$

This inequality is usually called **Kraft inequality**.

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- 2 Given a set of codeword lengths $\{l_1, \dots, l_m\}$, there exists a binary prefix code with these codeword lengths if and only if $l_i, i = 1, \dots, m$, satisfy the Kraft inequality

$$\sum_{i=1}^m 2^{-l_i} \leq 1.$$

0	00	000	0000
			0001
		001	0010
			0011
	01	010	0100
			0101
		011	0110
			0111
1	10	100	1000
			1001
		101	1010
			1011
	11	110	1100
			1101
		111	1110
			1111

0	00	000	0000
			0001
		001	0010
			0011
	01	010	0100
			0101
		011	0110
			0111
1	10	100	1000
			1001
		101	1010
			1011
	11	110	1100
			1101
		111	1110
			1111

0	00	000	0000
			0001
		001	0010
			0011
	01	010	0100
			0101
		011	0110
			0111
1	10	100	1000
			1001
		101	1010
			1011
	11	110	1100
			1101
		111	1110
			1111

0	00	000	0000
			0001
		001	0010
			0011
	01	010	0100
			0101
		011	0110
			0111
1	10	100	1000
			1001
		101	1010
			1011
	11	110	1100
			1101
		111	1110
			1111

0	00	000	0000
			0001
		001	0010
			0011
	01	010	0100
			0101
		011	0110
			0111
1	10	100	1000
			1001
		101	1010
			1011
	11	110	1100
			1101
		111	1110
			1111

- ★ We want to minimize the expected length code $L(C, X)$
- ★ The entropy is a lower bound:

$$L(C, X) \geq H(X)$$

- ★ **Optimal source codelengths:** $L(C, X)$ is minimized and is equal to $H(X)$ only if the codelengths are equal to the *Shannon information content*:

$$l_i = \log_2(1/p_i)$$

- ★ **Source coding theorem:** For a source (random variable) X , there exists a prefix code C with expected length satisfying:

$$H(X) \leq L(C, X) \leq H(X) + 1.$$