Coding theory - theory of error correcting codes - is one of the most interesting and applied part of informatics. Goals of coding theory are to develop systems and methods that allow to detect/correct errors caused when information is transmitted through noisy channels. All real communication systems that work with digitally represented data, as CD players, TV, fax machines, internet, satellites, mobiles, require to use error correcting codes because all real channels are, to some extent, noisy – due to various interference/destruction caused by the environment. Coding theory problems are therefore among the very basic and most frequent problems of storage and transmission of information. Coding theory results allow to create reliable systems out of unreliable systems to store and/or to transmit information. Coding theory methods are often elegant applications of very basic concepts and methods of (abstract) algebra.

This first chapter presents and illustrates the very basic problems, concepts, methods and results of coding theory.

**Coding - Basic Concepts**

Without coding theory and error-correcting codes there would be no deep-space travel and pictures, no satellite TV, no compact disc, no . . . no . . . no . . . .

Error-correcting codes are used to correct messages when they are (erroneously) transmitted through noisy channels.

A code \( C \) over an alphabet \( \Sigma \) is a subset of \( \Sigma^*(C \subseteq \Sigma^*) \). A q-nary code is a code over an alphabet of q-symbols. A binary code is a code over the alphabet \{0, 1\}.

Examples of codes

- \( C_1 = \{00, 01, 10, 11\} \)
- \( C_2 = \{000, 010, 101, 100\} \)
- \( C_3 = \{00000, 01101, 10111, 11011\} \)

**Channel**

is any physical medium in which information is stored or through which information is transmitted. (Telephone lines, optical fibres and also the atmosphere are examples of channels.)

Noise may be caused by sunspots, lighting, meteor showers, random radio disturbance, poor typing, poor hearing, . . . .

**Transmission Goals**

- Fast encoding of information.
- Similar messages should be encoded very differently
- Easy transmission of encoded messages.
- Fast decoding of received messages.
- Reliable correction of errors introduced in the channel.
- Maximum transfer of information per unit time.

**Basic Method of Fighting Errors: Redundancy!!!**

Example: 0 is encoded as 00000 and 1 is encoded as 11111.
Formally, a channel is described by a triple $C = (\Sigma, \Omega, p)$, where
- $\Sigma$ is an input alphabet
- $\Omega$ is an output alphabet
- $p$ is a probability distribution on $\Sigma \times \Omega$ and for $i \in \Sigma$, $o \in \Omega$, $p(i, o)$ is the probability that the output of the channel is $o$ if the input is $i$.

**IMPORTANT CHANNELS**

- **Binary symmetric channel** maps, with probability $p_0$ each binary input into the opposite one. Therefore, $Pr(0,1) = Pr(1,0) = p_0$ and $Pr(0,0) = Pr(1,1) = 1 - p_0$.
- **Binary erasure channel** with probability $p_0$, maps binary inputs into outputs in $\{0, 1, e\}$, where $e$ is so called the erasure symbol, and $Pr(0,0) = Pr(1,1) = Pr(p_0)$, $Pr(0,e) = Pr(1,e) = 1 - p_0$.
- **White noise Gaussian channel** that models errors in deep space.

**WHY WE NEED TO IMPROVE ERROR-CORRECTING CODES**

When error correcting capabilities of some code are improved - that is a better code is found - this has the following impacts:
- For the same quality of of received information it is possible to achieve that the transmission system operates in more severe conditions;
- For example:
  - It is possible to reduce the size of antennas or solar panels and the weight of batteries;
  - In space systems such savings can be measured in hundred of thousands of dollars;
  - In mobile telephone systems, improving the code enables the operators to increase the potential number of users in each cell.
- Another field of applications of error-correcting codes is that of mass memories: computer hard drives, CD-ROMs, DVDs and so on.

**BASIC CHANNEL CODING PROBLEMS**

**Summary:** The task of a channel coding is to encode the information sent over a communication channel in such a way that in the presence of some channel noise, errors can be detected and/or corrected.

There are two basic coding methods
- **BEC (Backward Error Correction)** Coding allows the receiver only errors detection. If an error is detected the sender is requested to retransmit the message.
- **FEC (Forward Error Correction)** Coding allows the receiver to correct a certain amount of errors.

**WHY WE NEED TO IMPROVE ERROR-CORRECTING CODES**

When error correcting capabilities of some code are improved - that is a better code is found - this has the following impacts:

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- Another field of applications of error-correcting codes is that of mass memories: computer hard drives, CD-ROMs, DVDs and so on.
The basic idea of so called majority voting decoding/principle or of maximal likelihood decoding/principle is to decode a received message \( w' \) by a codeword \( w \) that is the closest one to \( w' \) in the whole set of the potential codewords of a given code \( C \).

EXAMPLE

In case the encoding

\[
\begin{align*}
0 &\rightarrow 000 \\
1 &\rightarrow 111
\end{align*}
\]

is used, then the probability of the bit error is \( p \leq \frac{1}{2} \), and the following majority voting decoding

\[
\begin{align*}
000, 001, 010, 100 &\rightarrow 000 \\
111, 110, 101, 011 &\rightarrow 111
\end{align*}
\]

is used, then the probability of an erroneous decoding (for the case of 2 or 3 errors) is

\[
3p^2(1-p) + p^3 < p
\]

EXAMPLE: Coding of a path avoiding an enemy territory

Story Alice and Bob share an identical map (Fig. 1) gridded as shown in Fig.1. Only Alice knows the route through which Bob can reach her avoiding the enemy territory. Alice wants to send Bob the following information about the safe route he should take.

\[
\text{NNWNNWSSWWNNNNWWN}
\]

Three ways to encode the safe route from Bob to Alice are:

1. \( C_1 = \{N = 00, W = 01, S = 11, E = 10\} \)

   In such a case any error in the code word
   
   \[
   0000100000101111010100000000010100
   \]
   
   would be a disaster.

2. \( C_2 = \{000, 011, 101, 110\} \)

   A single error in encoding each of symbols N, W, S, E can be detected.

3. \( C_3 = \{00000, 01101, 10110, 11011\} \)

   A single error in decoding each of symbols N, W, S, E can be corrected.

BASIC TERMINOLOGY

Datawords - words of a message
Codewords - words of some code.
Block code - a code with all codewords of the same length.

Basic assumptions about channels

- **Code length preservation** Each output word of a channel has the same length as the input codeword.
- **Independence of errors** The probability of any one symbol being affected in transmissions is the same.

Basic strategy for decoding

For decoding we use the so-called maximal likehood principle, or nearest neighbor decoding strategy, or majority voting decoding strategy which says that the receiver should decode a received word \( w' \) as

the codeword \( w \) that is the closest one to \( w' \).
HAMMING DISTANCE

The intuitive concept of "closeness" of two words is well formalized through Hamming distance $h(x, y)$ of words $x, y$. For two words $x, y$

$$h(x, y) = \text{the number of symbols in which the words } x \text{ and } y \text{ differ.}$$

**Example:**

$$h(10101, 01100) = 3,$$  
$$h(\text{fourth, eighth}) = 4$$

**Properties of Hamming distance**

- $h(x, y) = 0 \iff x = y$
- $h(x, y) = h(y, x)$
- $h(x, z) \leq h(x, y) + h(y, z)$ (triangle inequality)

An important parameter of codes $C$ is their minimal distance.

$$h(C) = \min\{h(x, y) | x, y \in C, x \neq y\}.$$ Therefore, $h(C)$ is the smallest number of errors that can change one codeword into another.

**Basic error correcting theorem**

- A code $C$ can detect up to $s$ errors if $h(C) \geq s + 1$.
- A code $C$ can correct up to $t$ errors if $h(C) \geq 2t + 1$.

**Proof**

(1) Trivial.  (2) Suppose $h(C) \geq 2t + 1$. Let a codeword $x$ is transmitted and a word $y$ is received with $h(x, y) \leq t$. If $x' \neq x$ is a codeword, then $h(y, x') \geq t + 1$ because otherwise $h(y, x') < t + 1$ and therefore $h(x, y) \leq h(x, y) + h(y, x') < 2t + 1$ what contradicts the assumption $h(C) \geq 2t + 1$.

BINARY SYMMETRIC CHANNEL

Consider a transition of binary symbols such that each symbol has probability of error $p < \frac{1}{2}$.

The intuitive concept of "closeness" of two words is well formalized through Hamming distance $h(x, y)$ of words $x, y$. For two words $x, y$

$$h(x, y) = \text{the number of symbols in which the words } x \text{ and } y \text{ differ.}$$

**Example**

Let all $2^{11}$ binary words of length 11 be codewords and let the probability of a bit error be $p = 10^{-8}$. Let bits be transmitted at the rate $10^7$ bits per second. The probability that a word is transmitted incorrectly is approximately

$$11p(1 - p)^{10} \approx \frac{11}{10^7}.$$ Therefore $\frac{11}{10^7} \cdot \frac{10^7}{11} = 0.1$ of words per second are transmitted incorrectly.

Therefore, one wrong word is transmitted every 10 seconds, 360 erroneous words every hour and 8640 words every day without being detected.

Let now one parity bit be added. Any single error can be detected.!!

The probability of at least two errors is:

$$1 - (1 - p)^{12} - 12(1 - p)^{11}p \approx \binom{12}{1}(1 - p)^{10}p^2 \approx \frac{66}{10^9}.$$ Therefore, approximately $\frac{66}{10^9} \cdot \frac{10^7}{12} \approx 5.5 \cdot 10^{-9}$ words per second are transmitted with an undetectable error.

**Corollary** One undetected error occurs only every 2000 days! ($2000 \approx \frac{10^9}{5.5 \times 86400}$).

TWO-DIMENSIONAL PARITY CODE

The two-dimensional parity code arranges the data into a two-dimensional array and then to each row (column) parity bit is attached.

**Example**

Binary string

10001011000100101111

is represented and encoded as follows

1 0 0 0 1 1 0 0 0 0 1 0
0 1 1 0 0 0 1 0 0 1 0 0
0 1 0 0 1 0 1 1 1 0 1 0
0 0 1 1 1 1 1 0 1 0 1 1
0 1 1 1 1 1 0 1 0 1 1 0

**Question** How much better is two-dimensional encoding than one-dimensional encoding?

**Example**

Binary string

101010111000010111111

is received with $h' = 6$.

Consider a transition of binary symbols such that each symbol has probability of error $p$.

If $n$ symbols are transmitted, then the probability of $t$ errors is

$$p'(1 - p)^{n-t}\binom{n}{t}.$$ In the case of binary symmetric channels, the "nearest neighbour decoding strategy" is also "maximum likelihood decoding strategy".

**Example**

Consider $C = \{000, 111\}$ and the nearest neighbour decoding strategy. Probability that the received word is decoded correctly

as 000 is $(1 - p)^2 + 3p(1 - p)^2$,

as 111 is $(1 - p)^3 + 3p(1 - p)^2$.

Therefore, $P_{\text{err}}(C) = 1 - (1 - p)^3 + 3p(1 - p)^2$ is the probability of an erroneous decoding.

**Example**

If $p = 0.01$, then $P_{\text{err}}(C) = 0.000298$ and only one word in 3356 will reach the user with an error.
**NOTATIONS and EXAMPLES**

**Notation:** An \((n, M, d)\)-code \(C\) is a code such that
- \(n\) - is the length of codewords.
- \(M\) - is the number of codewords.
- \(d\) - is the minimum distance in \(C\).

**Example:**
- \(C_1 = \{00, 01, 10, 11\}\) is a \((2,4,1)\)-code.
- \(C_2 = \{000, 011, 101, 110\}\) is a \((3,4,2)\)-code.
- \(C_3 = \{00000, 01101, 10110, 11011\}\) is a \((5,4,3)\)-code.

**Comment:** A good \((n, M, d)\)-code has small \(n\), large \(M\) and also large \(d\).

---

**EXAMPLES from DEEP SPACE TRAVELS**

**Examples (Transmission of photographs from the deep space)**

- In 1965-69 **Mariner 4-5** probes took the first photographs of another planet - 22 photos. Each photo was divided into \(200 \times 200\) elementary squares - pixels. Each pixel was assigned 6 bits representing 64 levels of brightness. and so called **Hadamard code** was used.
  - Transmission rate was 8.3 bits per second.

- In 1970-72 **Mariners 6-8** took such photographs that each picture was broken into \(700 \times 832\) squares. So called **Reed-Muller \((32,64,16)\)** code was used.
  - Transmission rate was 16200 bits per second. (Much better quality pictures could be received)

---

**HADAMARD CODE**

In Mariner 5, 6-bit pixels were encoded using 32-bit long Hadamard code that could correct up to 7 errors.

**Hadamard code** has 64 codewords. 32 of them are represented by the \(32 \times 32\) matrix \(H = \{h_{ij}\}\), where \(0 \leq i, j \leq 31\) and

\[
h_{ij} = (-1)^{a_i b_j + a_i b_{j-1} + \ldots + a_0 b_0}
\]

where \(i\) and \(j\) have binary representations

\[
i = a_4 a_3 a_2 a_1 a_0, \quad j = b_4 b_3 b_2 b_1 b_0
\]

The remaining 32 codewords are represented by the matrix \(-H\).

Decoding was quite simple.

---

**CODES RATES**

For \(q\)-nary \((n, M, d)\)-code we define the code rate, or information rate, \(R\), by

\[
R = \frac{\log q}{n} M
\]

The code rate represents the ratio of the number of needed input data symbols to the number of transmitted code symbols.

If a \(q\)-nary code has code rate \(R\), then we say that it transmits \(R\) \(q\)-symbols per a channel use - or number of bits per channel use (bpc) in the case of binary alphabet.

Code rate \((6/32\) for Hadamard code\), is an important parameter for real implementations, because it shows what fraction of the communication bandwidth is being used to transmit actual data.
The ISBN-code I

Each book till 1.1.2007 had International Standard Book Number which was a 10-digit codeword produced by the publisher with the following structure:

<table>
<thead>
<tr>
<th>language</th>
<th>publisher</th>
<th>number</th>
<th>weighted check sum</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>07</td>
<td>709503</td>
<td>0</td>
</tr>
</tbody>
</table>

such that \[\sum_{i=1}^{10} (11 - i)x_i \equiv 0 \pmod{11}\]

The ISBN code was designed to detect: (a) any single error (b) any double error created by a transposition

Single error detection

Let \(X = x_1 \ldots x_{10}\) be a correct code and let \(Y = x_1 \ldots x_{j-1}y_jx_{j+1} \ldots x_{10}\) with \(y_j = x_j + a, a \neq 0\)

In such a case:

\[\sum_{i=1}^{11} y_i = \sum_{i=1}^{11} x_i + ja \neq 0 \pmod{11}\]

New ISBN code


New ISBN number can be obtained from the old one by preceding the old code with three digits 978.

For details about 13-digit ISBN see

http://www.en.wikipedia.org/Wiki/International_Standard_Book_Number

The ISBN-code II

Transposition detection

Let \(x_j \text{ and } x_k\) be exchanged.

\[\sum_{i=1}^{11} y_i = \sum_{i=1}^{11} x_i + (k - j)x_j + (j - k)x_k = (k - j)(x_j - x_k) \neq 0 \pmod{11}\]

if \(k \neq j\) and \(x_j \neq x_k\).

EQUIVALENCE of CODES

Definition Two \(q\)-ary codes are called equivalent if one can be obtained from the other by a combination of operations of the following type:

(a) a permutation of the positions of the code.
(b) a permutation of symbols appearing in a fixed position.

Question: Let a code be displayed as an \(M \times n\) matrix. To what correspond operations (a) and (b)?

Claim: Distances between codewords are unchanged by operations (a), (b).
 Consequently, equivalent codes have the same parameters \((n, M, d)\) (and correct the same number of errors).

Examples of equivalent codes

\[
\begin{pmatrix}
0 & 0 & 1 & 0 & 0 \\
0 & 0 & 0 & 1 & 1 \\
1 & 1 & 1 & 1 & 1 \\
1 & 1 & 0 & 0 & 0
\end{pmatrix}
\begin{pmatrix}
0 & 0 & 0 & 0 & 0 \\
0 & 1 & 1 & 0 & 1 \\
1 & 0 & 1 & 1 & 0 \\
1 & 1 & 0 & 1 & 1
\end{pmatrix}
\begin{pmatrix}
0 & 0 & 0 & 1 & 2 \\
1 & 1 & 1 & 2 & 0 \\
2 & 2 & 2 & 0 & 1
\end{pmatrix}
\]

Lemma Any \(q\)-ary \((n, M, d)\)-code over an alphabet \(\{0, 1, \ldots, q - 1\}\) is equivalent to an \((n, M, d)\)-code which contains the all-zero codeword \(00 \ldots 0\).

Proof Trivial.
THE MAIN CODING THEORY PROBLEM

A good \((n, M, d)\)-code should have a small \(n\), large \(M\) and large \(d\).

The main coding theory problem is to optimize one of the parameters \(n\), \(M\), \(d\) for given values of the other two.

**Notation:** \(A_q(n, d)\) is the largest \(M\) such that there is an \(q\)-nary \((n, M, d)\)-code.

**Theorem**

(a) \(A_q(n, 1) = q^n\);

(b) \(A_q(n, n) = q^2\).

**Proof**

(a) First claim is obvious;

(b) Let \(C\) be an \(q\)-nary \((n, M, n)\)-code. Any two distinct codewords of \(C\) have to differ in all \(n\) positions. Hence symbols in any fixed position of \(M\) codewords have to be different. Therefore \(A_q(n, n) \leq q\). Since the \(q\)-nary repetition code is \((n, q, n)\)-code, we get \(A_q(n, n) \geq q\).

**EXAMPLE**

**Example Proof that** \(A_2(5, 3) = 4\).

(a) Code \(C_3\) is a \((5, 4, 3)\)-code, hence \(A_2(5, 3) \geq 4\).

(b) Let \(C\) be a \((5, M, 3)\)-code with \(M = 5\).

- By previous lemma we can assume that 00000 \(\in C\).
- \(C\) has to contain at most one codeword with at least four 1's. (otherwise \(d(x, y) \leq 2\) for two such codewords \(x, y\))
- Since 00000 \(\in C\), there can be no codeword in \(C\) with at most one or two 1.
- Since \(d = 3\), \(C\) cannot contain three codewords with three 1's.
- Since \(M \geq 4\), there have to be in \(C\) two codewords with three 1's. (say 11100, 00111), the only possible codeword with four or five 1's is then 11011.

**DESIGN of ONE CODE from ANOTHER ONE**

**Theorem** Suppose \(d\) is odd. Then a binary \((n, M, d)\)-code exists iff a binary \((n + 1, M, d + 1)\)-code exists.

**Proof** Only if case: Let \(C\) be a binary \((n, M, d)\) code. Let

\[
C' = \{x_1 \ldots x_n | x_1 \ldots x_n \in C, x_{n+1} = \left(\sum_{i=1}^{n} x_i \right) \mod 2\}
\]

Since parity of all codewords in \(C'\) is even, \(d(x', y')\) is even for all \(x', y' \in C'\).

Hence \(d(C')\) is even. Since \(d \leq d(C') \leq d + 1\) and \(d\) is odd,
\[
d(C') = d + 1.
\]

Hence \(C'\) is an \((n + 1, M, d + 1)\)-code.

If case: Let \(D\) be an \((n + 1, M, d + 1)\)-code. Choose code words \(x, y\) of \(D\) such that \(d(x, y) = d + 1\).

Find a position in which \(x, y\) differ and delete this position from all codewords of \(D\).

Resulting code is an \((n, M, d)\)-code.

**A COROLLARY**

**Corollary:**

If \(d\) is odd, then \(A_2(n, d) = A_2(n + 1, d + 1)\).

If \(d\) is even, then \(A_2(n, d) = A_2(n - 1, d - 1)\).

**Example**

\(A_2(5, 3) = 4 \Rightarrow A_2(6, 4) = 4\)

\((5, 4, 3)\)-code \(\Rightarrow (6, 4, 4)\)-code

\[
\begin{array}{cccc}
0 & 0 & 0 & 0 \\
0 & 1 & 1 & 0 \\
1 & 0 & 1 & 1 \\
1 & 1 & 0 & 1 \\
\end{array}
\]

by adding check.
A SPHERE and its CONTENTS

Notation $F_q^n$ is a set of all words of length $n$ over the alphabet $\{0, 1, 2, \ldots, q - 1\}$.

Definition For any codeword $u \in F_q^n$ and any integer $r \geq 0$ the sphere of radius $r$ and centre $u$ is denoted by

$$S(u, r) = \{v \in F_q^n | h(u, v) \leq r\}.$$  

Theorem A sphere of radius $r$ in $F_q^n$, $0 \leq r \leq n$ contains

$${n \choose r} + \sum_{j=0}^{r} j(q-1)^j$$

words.

Proof Let $u$ be a fixed word in $F_q^n$. The number of words that differ from $u$ in $m$ positions is

$${n \choose m}(q - 1)^m.$$  

A GENERAL UPPER BOUND on $A_q(n, d)$

Example An $(7, M, 3)$-code is perfect if

$$M \left( {7 \choose 3} + {7 \choose 2} \right) = 2^7$$

i.e. $M = 16$

An example of such a code:

$$C4 = \{0000000, 1111111, 1000101, 1100010, 0110001, 1011000, 0101100, 0010110, 0001011, 0111010, 0011110, 1001110, 0100111, 1001011, 1101011, 1110100\}$$

Table of $A_2(n, d)$ from 1981

<table>
<thead>
<tr>
<th>$n$</th>
<th>$d = 3$</th>
<th>$d = 5$</th>
<th>$d = 7$</th>
</tr>
</thead>
<tbody>
<tr>
<td>5</td>
<td>4</td>
<td>2</td>
<td>-</td>
</tr>
<tr>
<td>6</td>
<td>8</td>
<td>2</td>
<td>-</td>
</tr>
<tr>
<td>7</td>
<td>16</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>8</td>
<td>20</td>
<td>4</td>
<td>2</td>
</tr>
<tr>
<td>9</td>
<td>40</td>
<td>6</td>
<td>2</td>
</tr>
<tr>
<td>10</td>
<td>72-79</td>
<td>12</td>
<td>2</td>
</tr>
<tr>
<td>11</td>
<td>144-158</td>
<td>24</td>
<td>4</td>
</tr>
<tr>
<td>12</td>
<td>256</td>
<td>32</td>
<td>4</td>
</tr>
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<td>512</td>
<td>64</td>
<td>8</td>
</tr>
<tr>
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<td>1024</td>
<td>128</td>
<td>16</td>
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<tr>
<td>15</td>
<td>2048</td>
<td>256</td>
<td>32</td>
</tr>
<tr>
<td>16</td>
<td>2560-3276</td>
<td>256-340</td>
<td>36-37</td>
</tr>
</tbody>
</table>

For current best results see [http://www.codetables.de](http://www.codetables.de)

LOWER BOUND for $A_q(n, d)$

The following lower bound for $A_q(n, d)$ is known as Gilbert-Varshamov bound:

Theorem Given $d \leq n$, there exists a $q$-ary $(n, M, d)$-code with

$$M \geq \frac{q^n}{\sum_{j=0}^{d-1} {n \choose j}(q-1)^j}$$

and therefore

$$A_q(n, d) \geq \frac{q^n}{\sum_{j=0}^{d-1} {n \choose j}(q-1)^j}$$
ERROR DETECTION

Error detection is much more modest aim than error correction. Error detection is suitable in the cases that channel is so good that probability of an error is small and if an error is detected, the receiver can ask the sender to renew the transmission.

For example, two main requirements for many telegraphy codes used to be:
- Any two codewords had to have distance at least 2;
- No codeword could be obtained from another codeword by transposition of two adjacent letters.

PICTURES of SATURN TAKEN by VOYAGER

Pictures of Saturn taken by Voyager, in 1980, had $800 \times 800$ pixels with 8 levels of brightness.

Since pictures were in color, each picture was transmitted three times; each time through different color filter. The full color picture was represented by

$$3 \times 800 \times 800 \times 8 = 13360000 \text{ bits}.$$  
To transmit pictures Voyager used the so called Golay code $G_{24}$.

GENERAL CODING PROBLEM

Important problems of information theory are how to define formally such concepts as information and how to store or transmit information efficiently.

Let $X$ be a random variable (source) which takes any value $x$ with probability $p(x)$. The entropy of $X$ is defined by

$$S(X) = - \sum_x p(x) \log p(x)$$

and it is considered to be the information content of $X$.

In a special case of a binary variable $X$ which takes on the value 1 with probability $p$ and the value 0 with probability $1 - p$

$$S(X) = H(p) = -p \log p - (1 - p) \log (1 - p)$$

Problem: What is the minimal number of bits needed to transmit $n$ values of $X$?

Basic idea: To encode more (less) probable outputs of $X$ by shorter (longer) binary words.

Example (Morse code - 1838)

<table>
<thead>
<tr>
<th>mess.</th>
<th>code</th>
</tr>
</thead>
<tbody>
<tr>
<td>0000</td>
<td>10</td>
</tr>
<tr>
<td>0001</td>
<td>001</td>
</tr>
<tr>
<td>0010</td>
<td>011</td>
</tr>
<tr>
<td>0011</td>
<td>1100</td>
</tr>
<tr>
<td>0100</td>
<td>1101</td>
</tr>
<tr>
<td>0101</td>
<td>1110</td>
</tr>
<tr>
<td>0110</td>
<td>1111</td>
</tr>
<tr>
<td>1000</td>
<td>11101</td>
</tr>
<tr>
<td>1001</td>
<td>11110</td>
</tr>
<tr>
<td>1010</td>
<td>11111</td>
</tr>
<tr>
<td>1011</td>
<td>111111</td>
</tr>
<tr>
<td>1100</td>
<td>1111101</td>
</tr>
<tr>
<td>1101</td>
<td>111111</td>
</tr>
<tr>
<td>1110</td>
<td>1111111</td>
</tr>
<tr>
<td>1111</td>
<td>11111111</td>
</tr>
</tbody>
</table>

Observe that this is a prefix code - no codeword is a prefix of another codeword.

SHANNON’s NOISELESS CODING THEOREM

Shannon’s noiseless coding theorem says that in order to transmit $n$ values of $X$, we need, and it is sufficient, to use $nS(X)$ bits.

More exactly, we cannot do better than the bound $nS(X)$ says, and we can reach the bound $nS(X)$ as close as desirable.

Example Let a source $X$ produce the value 1 with probability $p = \frac{1}{4}$ and the value 0 with probability $1 - p = \frac{3}{4}$

Assume we want to encode blocks of the outputs of $X$ of length 4.

By Shannon’s theorem we need $4H(\frac{1}{4}) = 3.245$ bits per blocks (in average)

A simple and practical method known as Huffman code requires in this case 3.273 bits per a 4-bit message.

Notation $\log(n)$ [log] will be used for binary, natural and decimal logarithms.
The subject of error-correcting codes arose originally as a response to practical problems in the reliable communication of digitally encoded information.

The discipline was initiated in the paper


Shannon’s paper started the scientific discipline information theory and error-correcting codes are its part.

Originally, information theory was a part of electrical engineering. Nowadays, it is an important part of mathematics and also of informatics.

In the introduction to his seminal paper “A mathematical theory of communication” Shannon wrote:

The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point.
At the beginning of this chapter the process encoding-channel transmission-decoding was illustrated as follows:

In that process a binary message is at first encoded into a binary codeword, then transmitted through a noisy channel, and, finally, the decoder received for decoding a potentially erroneous binary message and made an error correction.

This is a simplified view of the whole process. In practice the whole process looks quite differently.

Here is a more realistic view of the whole encoding-transmission-decoding process:

that is
- a binary message is at first transferred to a binary codeword;
- the binary codeword is then transferred to an analogous signal;
- the analogous signal is then transmitted through a noisy channel;
- the received analogous signal is then transferred to a binary form that can be used then for decoding and decoding takes place.

In case the analogous noisy signal is transferred before decoding to the binary signal we talk about hard decoding;
In case the output of analogous-digital deoding is a pair \((p_b, b)\) where \(p_b\) is the probability that the output is the bit \(b\) (or a weight of such a binary output (often given by a number from an interval \((-V_{\text{max}}, V_{\text{max}})\)), we talk about soft decoding.

A very important case in practise, especially for space communication, is so-called additive white Gaussian noise (AWGN) and the channel with such a noise is called Gaussian channel.
HARD versus SOFT DECODING - COMMENTS

When the signal received by the decoder comes from a device capable of producing estimations of an analogue nature on the binary transmitted data the error correction capability of the decoder can greatly be improved.

Since the decoder has in such a case information about reliability of data received decoding on the basis of finding the codeword with minimal Hamming distance does not have to be optimal and optimal decoding may depend on the type of noise involved.

For example, in an important practical case of the Gaussian white noise one search at the minimal likelihood decoding for a codeword with minimal Euclidean distance.

BASIC FAMILIES of CODES

Two basic families of codes are

Block codes called also as algebraic codes that are appropriate to encode blocks of data of the same length and independent one from the other. Their encoders have often a huge number of internal states and decoding algorithms are based on techniques specific for each code.

Stream codes called also as convolution codes that are used to protect continuous flows of data. Their encoders often have only small number of internal states and then decoders can use a complete representation of states using so called trellises, iterative approaches via several simple decoders and an exchange of information of probabilistic nature.

Hard decoding used to be mainly used for block codes and soft one for stream codes. However, distinctions between these two families of codes are tending to blur.

NOTATIONAL COMMENT

The term code is often used also to denote a specific encoding algorithm that transfers any dataword, say of the size $k$, into a codeword, say of the size $n$. The set of all such codewords then forms the code in the original sense.

For the same code there can be many encoding algorithms that map the same set of datawords into different codewords.

SYSTEMATIC CODES I

A code is called systematic if its encoder transmit a message (an input dataword) $w$ into a codeword of the form $wc_w$, or $(w, c_w)$. That is if the codeword for the dataword $w$ consists of two parts: dataword $w$ (called also information part) and redundancy part $c_w$.

Nowadays most of the stream codes that are used in practice are systematic.

An example of a systematic encoder, that produces so called extended Hamming $(8, 4, 1)$ code is in the following figure.
That is the encoder

where

\[ r_j = m_j \oplus \bigoplus_{i=1}^{3} m_i \quad j \in \{0, 1, 2, 3\} \]

A use of the ML principle for decoding amounts to searching for the output

\[ m'_0 m'_1 m'_2 m'_3 r'_0 r'_1 r'_2 r'_3 \]

of such a codeword \( m_0 m_1 m_2 m_3 r_0 r_1 r_2 r_3 \) that minimizes the expression

\[ \sum_{j=0}^{3} m_j \oplus m'_j + \sum_{j=0}^{3} r_j \oplus r'_j \]

The decoder that is capable of doing that is called ML-decoder.

In his telegraphs Morse used the following two-character audio alphabet

- **TIT** or dot — a short tone lasting four hundredths of second;
- **TAT** or dash — a long tone lasting twelve hundredths of second.

Morse could called these characters as 0 and 1

The binary elements 0 and 1 were first called **bits** by J. W. Tuckley in 1943.