

# Introduction to coding theory

CoCoNut, 2016  
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Steven Roman, Introduction to coding and information theory, Springer, *Undergraduate texts in mathematics*, 1997

# What does she say?

“Wel\*ome to t\*is c\*ass!” →

# What does she say?

“Wel\*ome to t\*is c\*ass!” → “Welcome to this class!”

Why is this example working?

- English has in built **redundancy**, so that it can tolerate *errors*.

# Coding theory I

More in general, consider the following applications of *data storage* or *transmission*:

- CDs and DVDs
- Satellite/Digital Television
- Deep space probes
- Internet communications
- Mobile phones
- Computer hard disks/memory/floppy etc

In all of these the data can become corrupted.

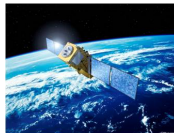
- It is prone to **errors**

However they still work

- **How?**

# Coding theory - Applications

- Internet
- Mobile phones
- Satellite broadcast
  - TV
- Deep space telecommunications
  - Mars Rover
- Data storage



Codes are all around us!

# Coding theory - The birth

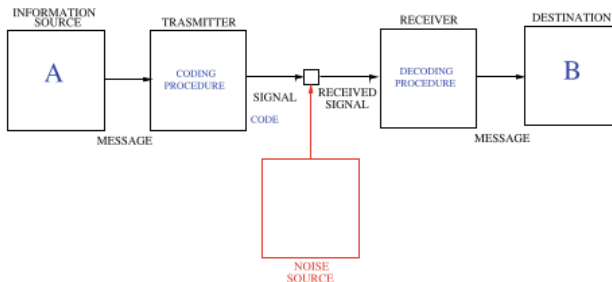
*“The fundamental problem of communication is that of reproducing at one point either exactly or approximately a message selected at another point”*  
(Claude Shannon, 1948)

- In 1948, Claude E. Shannon wrote “*A Mathematical Theory of Communication*”, which marked the beginning of both Information and Coding Theory
- In 1950, Richard W. Hamming wrote “*Error Detecting and Error Correcting Codes*”, which was the first paper explicitly introducing error-correcting codes



# Coding theory II

The general idea is that of adding some kind of redundancy to the message that we want to send over a communication channel





# Digital Data

Digital data is sent as a series of ones and zeros.

- 11110101111101010100011010101011

Sometimes an error occurs:

- 111101**1**1111101010100011010101011

We would like to be able to either detect or correct such errors.

## Detection

- Good if we can request a resend of the data

## Correction

- Needed if data cannot be resent (e.g. CD/DVD) or too costly to resend (e.g. deep space probe)

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# Simple Error Detection

Most data is first bundled up into a group of bits before sending

- e.g. 4, 8, 32 or 64 bits at a time

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A simple detection trick is to add a **parity bit**

Suppose we wish to transmit 4 bits

- 0110

We add in an extra bit which signals whether the original data

has an **even** or **odd** number of ones

- The extra bit denotes the **parity** of the original bits

0110 → 0110**0**

1111 → 1111**0**

1000 → 1000**1**

1011 → 1011**1**

# Simple Error Detection

The previous example can be described mathematically as follows.

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We wish to send four message bits

$$m_1 m_2 m_3 m_4 \in \{0, 1\}^4$$

To do this we add a fifth bit equal to

$$m_5 = m_1 \oplus m_2 \oplus m_3 \oplus m_4$$

where

$$x \oplus y = x + y \pmod{2}$$

The resulting five bits is called a **codeword**

# Simple Error Detection

We can now detect whether a **single** error has occurred.

Suppose you receive the following data using the previous example:

- 10101    **Errors**
- 01110    **Errors**
- 11101    **No errors**
- 11111    **Errors**
- 00000    **No errors**
- 00001    **Errors**

Trouble is we do not know where the errors occurred

# Detecting errors - Hamming code I

Again sticking to four bits of message

$$m_1 m_2 m_3 m_4$$

The idea is to use multiple parity-check bits.

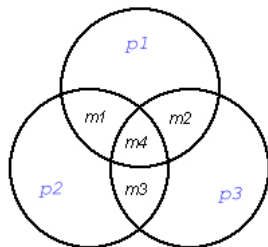
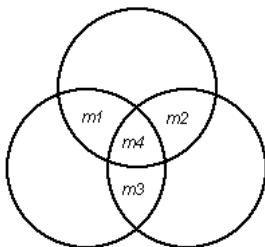


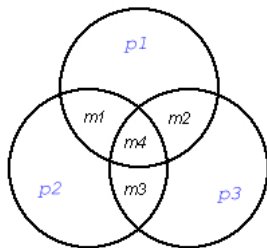
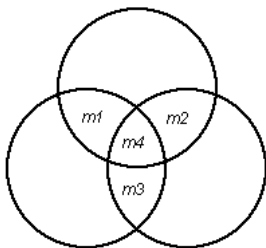
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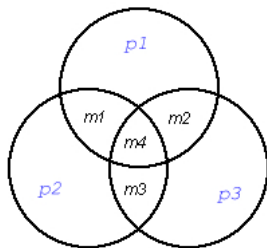
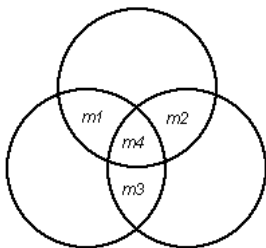
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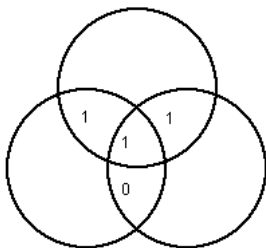
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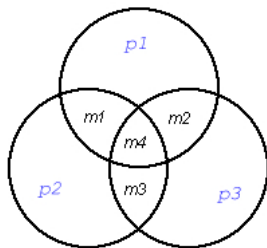
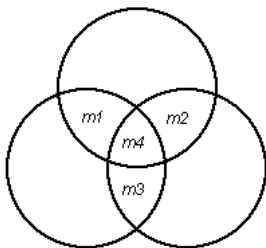




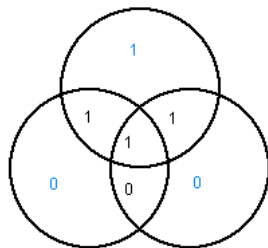
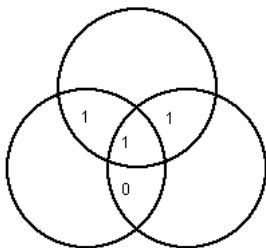


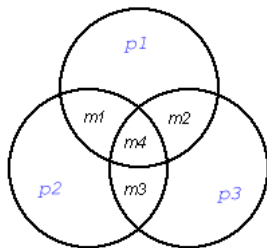
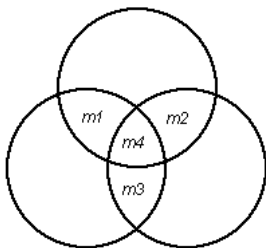
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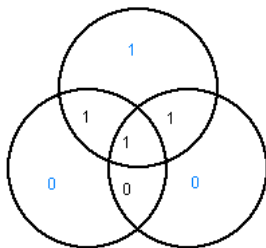
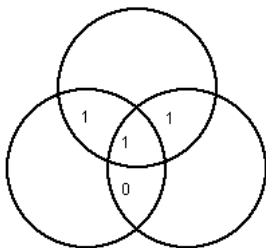


Suppose  $\mathbf{m} = 1101$  is the message





Suppose  $\mathbf{m} = 1101$  is the message  $\rightarrow 1101100$  is the codeword

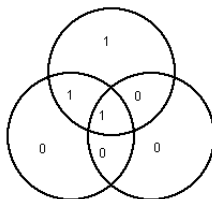


## Detecting errors - Hamming code III

Suppose that after transmission one symbol is flipped and  $\mathbf{r} = 1001100$  is received.

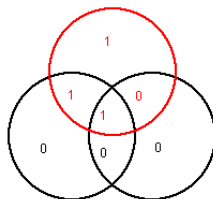
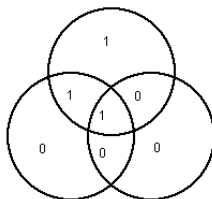
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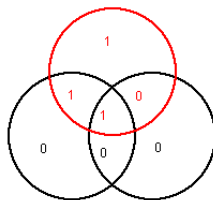


$1 + 0 + 1 + 1 = 1$  NOT OK!

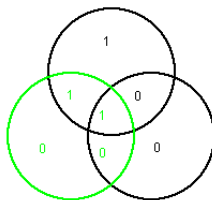
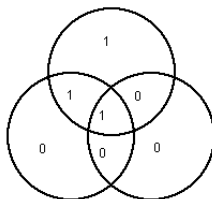


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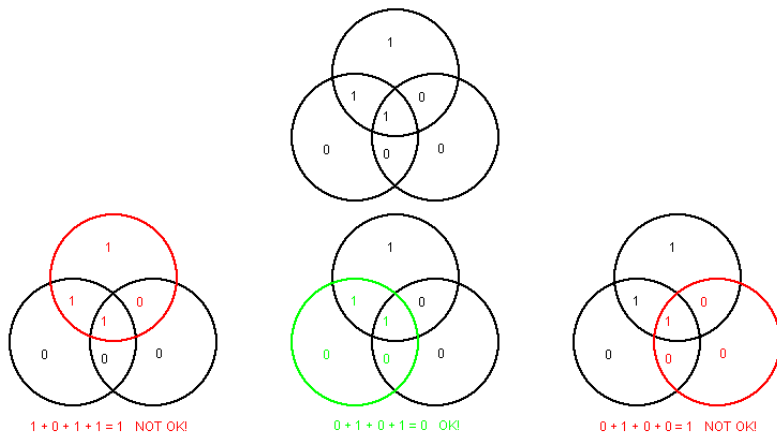
$$1 + 0 + 1 + 1 = 1 \text{ NOT OK!}$$



$$0 + 1 + 0 + 1 = 0 \text{ OK!}$$

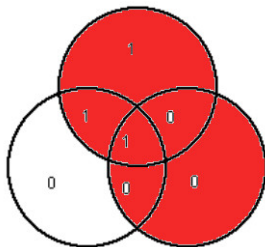
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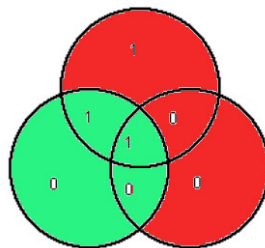
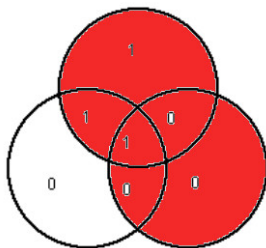
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$\mathbf{c} = 1101100$  and  $\mathbf{r} = 1001100$



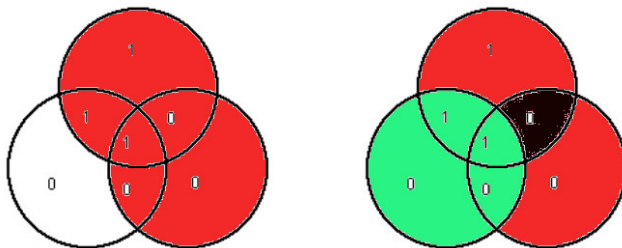
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# Correcting errors - Hamming code IV

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→ the error is at  $m_2$

# Hamming code - Basic idea

- Use multiple parity bits, each covering a subset of the message bits

$$m_1 + m_4 + m_2 + p_1 = 0 \longrightarrow \{m_1, m_4, m_2, p_1\} = C_1$$

$$m_1 + m_3 + m_4 + p_2 = 0 \longrightarrow \{m_1, m_3, m_4, p_2\} = C_2$$

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- No two message bits belong to exactly the same subsets. In this way it is possible to **correct** one error and to **detect** two errors.

# Hamming code - Basic idea

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$$m_1 + m_4 + m_2 + p_1 = 0 \longrightarrow \{m_1, m_4, m_2, p_1\} = C_1 \text{ OK}$$

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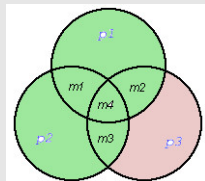
$$m_2 + m_4 + m_3 + p_3 = 0 \longrightarrow \{m_2, m_4, m_3, p_3\} = C_3 \text{ ERRORS}$$

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- No two message bits belong to exactly the same subsets. In this way it is possible to **correct** one error and to **detect** two errors.

## Example

Suppose two errors occurred, at  $m_2$  and  $p_1$ . Then:

- The code detects that some errors occurred
- The code concludes the error is at  $p_3$ , introducing an extra error



# Hamming code

◀ Return

- Enc :  $\{0,1\}^4 \rightarrow \{0,1\}^7$  that maps the  $2^4$  strings of 4 bits **m** into a **codeword c**
- We can write down all the codewords:

Information bits	Codeword	Information bits	Codeword
0000	0000000	1000	1000110
0001	0001111	1001	1001001
0100	0010101	1010	1010101
0011	0011100	1011	1011010
0010	0010011	1100	1100011
0101	0101010	1101	1101100
0110	0110110	1110	1110000
0111	0111001	1111	1111111

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0101	0101010	1101	1101100
0110	0110110	1110	1110000
0111	0111001	1111	1111111

- $C$  contains 16 codewords of **length 7**

# Block codes - Notation I

- Let  $\mathcal{A}$  be an alphabet of cardinality  $q$



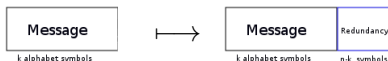
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- We consider codes  $C$  over  $\mathcal{A}$ . If  $q = 2$  the code is called **binary**
- Let  $\text{Enc}$  be an injective map:

$$\text{Enc} : \mathcal{A}^k \longrightarrow \mathcal{A}^n \quad C \text{ is the image of Enc}$$



the entire block is called **codeword**

- $n$  is the length of a codeword

## Definition

A **block code** is a code with fixed length  $n$ , i.e. a non-empty subset of  $\mathcal{A}^n$

- If a block code  $C \subseteq \mathcal{A}^n$  contains  $M = q^k$  codewords, then  $M$  is the **size** of  $C$

# Block codes - Notation II

A block code of length  $n$  and size  $M$  is denoted by  $(n, M)$ -code

- $k = \log_q(M)$     message length

- $n - \log_q(M)$     redundancy

- $R = \frac{\log_q(M)}{n}$     information rate

Average amount of real information in each block of  $n$  symbols transmitted over a channel

► Hamming

## Example

The Hamming code we have seen before is a binary  $(7, 16)$  block code with information rate  $4/7$ .

## Definition (Hamming distance)

Given two strings  $\mathbf{x}$  and  $\mathbf{y} \in \mathcal{A}^n$ , the **Hamming distance** between  $\mathbf{x}$  and  $\mathbf{y}$  is

$$d(\mathbf{x}, \mathbf{y}) = |\{i | x_i \neq y_i\}|.$$

### Example

 $\mathbf{v}_1 = 01011$  $\mathbf{v}_2 = 11110$ 

$$d(\mathbf{v}_1, \mathbf{v}_2) = 3$$

 $\mathbf{w}_1 = 3211$  $\mathbf{w}_2 = 0213$ 

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### Definition (Code distance)

The (Hamming) **minimum distance of a code**  $C$  is given by

$$d(C) = \min\{d(\mathbf{x}, \mathbf{y}) \mid \mathbf{x}, \mathbf{y} \in C, \mathbf{x} \neq \mathbf{y}\}.$$

### Definition (Hamming weight)

The **Hamming weight** of a string  $\mathbf{x}$ ,  $\text{wt}(\mathbf{x})$ , is defined as the number of non-zero symbols in the string.

# Decoding problem

Why is the distance of a code important?

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- ① if  $\mathbf{r} \in C$  then no correction is needed



- ② if  $\mathbf{r} \notin C$ , then some errors occurred



## Decoding problem - Why is $d(C)$ important?

If  $\mathbf{r} \notin C$ : the decoder has to find the codeword  $\mathbf{c}$  that has been sent



A possible strategy is the **Maximum Likelihood Decoding (MLD)**: find the most likely codeword transmitted, i.e. the codeword  $\mathbf{c}$  which maximizes the probability that  $\mathbf{r}$  is the received word given that  $\mathbf{c}$  has been sent.

We will see that for some types of channel MLD is equivalent to finding the coderword  $\mathbf{c}$  closest to  $\mathbf{r}$  in the Hamming distance (**Nearest neighbour decoding**):

$$\min_{\mathbf{c} \in C} d(\mathbf{r}, \mathbf{c})$$

From now on we will assume a type of channel such that we can use the minimum distance decoding to perform MLD

# Decoding problem - Why is $d(C)$ important?

Let  $\mathbf{x} \in \mathcal{A}^n$  and  $t \in \mathbb{N}$ , define

$$\mathcal{B}_t(\mathbf{x}) = \{\mathbf{y} \in \mathcal{A}^n \mid d(\mathbf{x}, \mathbf{y}) \leq t\}$$

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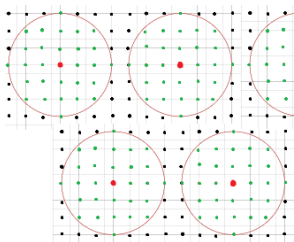
Image to cover the entire space  $\mathcal{A}^n$  of balls of radius  $\lfloor \frac{d-1}{2} \rfloor$  centered at distinct codewords:

# Decoding problem - Why is $d(C)$ important?

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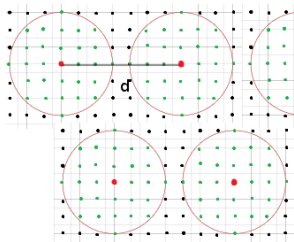


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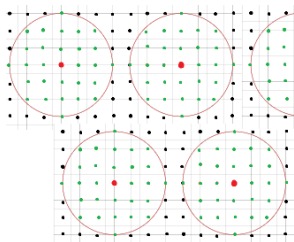
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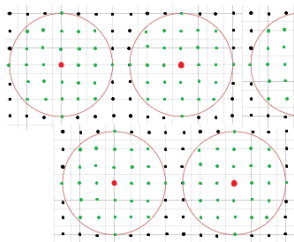
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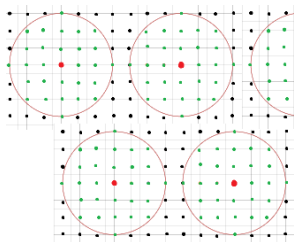


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- if  $\mathbf{r}$  = green word, we should correct it to the red codeword that is the center of the ball it lies in
- if  $\mathbf{r}$  = black word, then we are not able to correct because, if we increase the radii, balls would overlap

# Error correction and error detection capability

More formally, we have the following definition:

- The **error detection capability** of a code  $C$  is the number  $e$  of errors that the code can detect. A  $e$ -error detecting code has minimum distance  $d = e + 1$ .
- The **error correction capability** of a code  $C$  is the number of errors that the code can correct. A  $t$ -error detecting code has minimum distance  $d$  such that  $d = \lfloor \frac{t-1}{2} \rfloor$ .

# Erasures

In a similar way we can define the **erasure correction capability** of a code. An **erasure** occurs when a transmitted symbol is unreadable and at its place an extra symbol  $\epsilon$  is introduced.

## Example

$$\mathbf{c} = 1001100 \longrightarrow \mathbf{r} = 100\epsilon 100$$

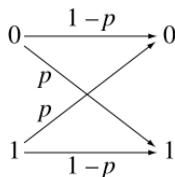
- A code can correct  $s$  erasures if  $s < d$
- The condition for simultaneous correction of  $t$  errors and  $s$  erasures is

$$d \geq 2t + s + 1.$$

# Binary Symmetric Channel (BSC)

$\mathcal{X} = \{0, 1\}$  input alphabet and  $\mathcal{Y} = \{0, 1\}$  output alphabet.

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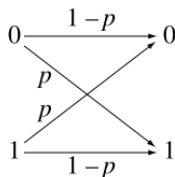
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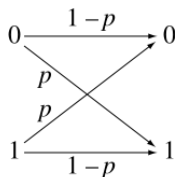
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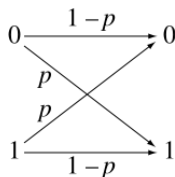
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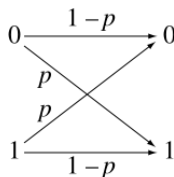
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- $\Pr(\mathbf{c}|\mathbf{c}) = (1 - p)^5$
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Suppose  $\mathbf{c}$  is transmitted codeword and  $\mathbf{r}$  is received word  $\rightarrow \mathbf{c} = \mathbf{r} + \mathbf{e}$

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Suppose  $\mathbf{c}$  is transmitted codeword and  $\mathbf{r}$  is received word  $\rightarrow \mathbf{c} = \mathbf{r} + \mathbf{e}$   
Given two codewords  $\mathbf{c}_1, \mathbf{c}_2$ , then

$$\begin{aligned}\Pr(\mathbf{r}|\mathbf{c}_1) \leq \Pr(\mathbf{r}|\mathbf{c}_2) &\iff d(\mathbf{r}, \mathbf{c}_1) \geq d(\mathbf{r}, \mathbf{c}_2) \\ &\iff \text{wt}(\mathbf{r} + \mathbf{c}_1) \geq \text{wt}(\mathbf{r} + \mathbf{c}_2) \\ &\iff \text{wt}(\mathbf{e}_1) \geq \text{wt}(\mathbf{e}_2)\end{aligned}$$

*The most likely codeword sent is the one corresponding to the error of smallest weight*

# Do we need more structure?

**Binary Hamming code (7, 16):**  $\text{Enc} : \{0, 1\}^4 \rightarrow \{0, 1\}^7$

Information bits	Codeword	Information bits	Codeword
0000	0000000	1000	1000110
0001	0001111	1001	1001001
0100	0010101	1010	1010101
0011	0011100	1011	1011010
0010	0010011	1100	1100011
0101	0101010	1101	1101100
0110	0110110	1110	1110000
0111	0111001	1111	1111111

We need  $n \cdot 2^k$  bits to store a binary code  $\text{Enc} : \{0, 1\}^k \rightarrow \{0, 1\}^n$

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We need extra structure that would facilitate a succinct representation of the code

# Can we do better?

Mathematically we can describe the  $(7, 16)_2$  Hamming code by a matrix

$$G = \begin{pmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{pmatrix},$$

so that, if we represent a message by the vector  $\mathbf{m} = (m_1 \ m_2 \ m_3 \ m_4)$ , we can encode by computing

$$\mathbf{c} = \mathbf{m} \cdot G$$

Suppose we wish to transmit  $\mathbf{m} = (1 \ 0 \ 1 \ 0)$ , we then compute

$$(1 \ 0 \ 1 \ 0) \cdot \begin{pmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{pmatrix} = (1 \ 0 \ 1 \ 0 \ 1 \ 0 \ 1)$$

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$$(1010) \cdot \begin{pmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{pmatrix} = (1010101)$$

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# Linear codes - Definition

The previous example is an example of **linear code**.

## Definition (Linear code)

Let  $q$  be a prime power. Then  $C \subseteq \{0, 1, \dots, q-1\}^n = \mathbb{F}_q^n$  is a linear code if it is a linear subspace of  $\mathbb{F}_q^n$ . If  $C$  has dimension  $k$  and distance  $d$  then it will be referred to as an  $[n, k, d]_q$  or just an  $[n, k]_q$  code.

- $\mathbb{F}_q^n$  denote the vector space of all  $n$ -tuples over the finite field  $\mathbb{F}_q$ .