



# *Coding Theory*

CO 331



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# Preface

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*Sibelius Peng*

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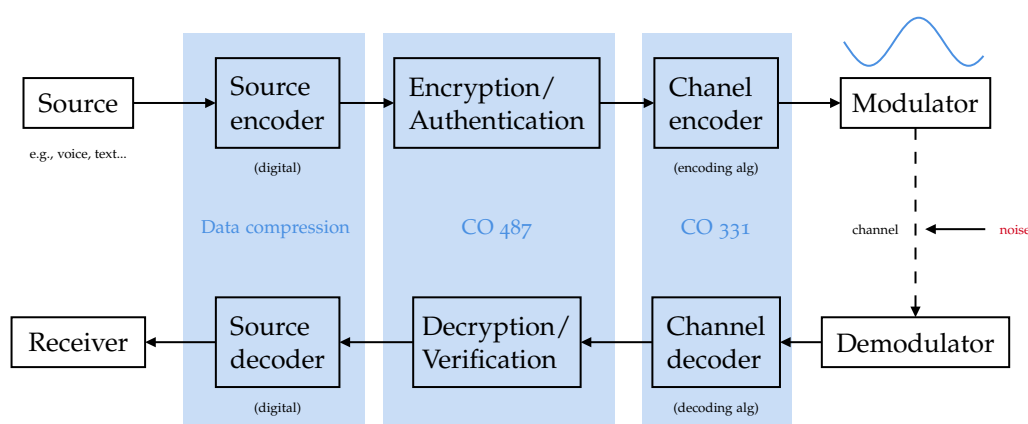
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# Introduction

Coding theory is about clever ways of adding redundancy to messages to allow (efficient) error detection and error correction.

Here is our communication model:



## Example: Parity Code

**Encoding algorithm** Add a 0 bit to the (binary) msg  $m$  if the number of 1's in  $m$  is even; else add a 1 bit.

**Decoding algorithm** If the number of 1's in a received msg  $r$  is even, then accept  $r$ ; else declare that an error has occurred.

## Example: Replication Code

Source msgs	Codeword	# err/codeword (always) detected	# err/codeword (always) corrected *	Information rate
0	0	0	0	1
1	1	0	0	1
0	00	1	0	$\frac{1}{2}$
1	11	1	0	$\frac{1}{2}$
0	000	2	1	$\frac{1}{3}$
1	111	2	1	$\frac{1}{3}$
0	0000	3	1	$\frac{1}{4}$
1	1111	3	1	$\frac{1}{4}$
0	00000	4	2	$\frac{1}{5}$
1	11111	4	2	$\frac{1}{5}$

encoding algorithm  
→

\*: using "nearest neighbour decoding"

## Goal of Coding Theory

Design codes so that:

1. High information rate
2. High error-correcting capability
3. Efficient encoding & decoding algorithms

## Course Overview

This course deals with *algebraic methods* for designing good (block) codes. The focus is on error correction (not on error detection). These codes are used in wireless communications, space probes, CD/DVD players, storage, QR codes, etc.

Some modern stuff are not covered: Turbo codes, LDPC codes, Raptor codes, ... Their math theories are not so elegant as algebraic codes.

## The big picture

Coding theory in its broadest sense deals with techniques for the *efficient, secure* and *reliable* transmission of data over communication channels that may be subject to *non-malicious errors* (noise) and *adversarial intrusion*. The latter includes passive intrusion (eavesdropping) and active intrusion (injection/deletion/modification).

# Fundamentals

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## 1.1 Basic Definitions and Concepts

### alphabet

An **alphabet**  $A$  is a finite set of  $q \geq 2$  symbols.

### word

A **word** is a finite sequence of symbols from  $A$  (also: vector, tuple).

### length

The **length** of a word is the number of symbols it has.

### code

A **code**  $C$  over  $A$  is a set of words (of size  $\geq 2$ ).

### codeword

A **codeword** is a word in the code  $C$ .

### block code

A **block code** is a code in which all codewords have the same length.

A **block code of length  $n$  containing  $M$  codewords over  $A$**  is a subset  $C \subseteq A^n$  with  $|C| = M$ .  $C$  is called an  $[n, M]$ -code over  $A$ .

**Example:**

$A = \{0, 1\}$ .  $C = \{00000, 11100, 00111, 10101\}$  is a  $[5, 4]$ -code over  $\{0, 1\}$ .

Messages		Codewords
00	→	00000
10	→	11100
01	→	00111
11	→	10101

Encoding of messages (1-1 map)

**Assumptions about the communications channel**

- (1) The channel only transmits symbols from  $A$  (“hard decision decoding”).
- (2) No symbols are deleted, added, interchanged or transposed during transmission.
- (3) The channel is a  $q$ -symmetric channel:

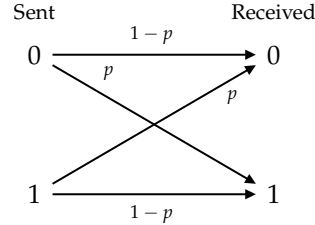
Let  $A = \{a_1, \dots, a_q\}$ . Let  $X_i$  = the  $i^{\text{th}}$  symbol sent. Let  $Y_i$  = the  $i^{\text{th}}$  symbol received. Then for all  $i \geq 1$ , and all  $i \leq j, k \leq q$ ,

$$\Pr(Y_i = a_j | X_i = a_k) = \begin{cases} 1 - p, & \text{if } j = k \\ \frac{p}{q-1}, & \text{if } j \neq k. \end{cases}$$

$p$  is called the **symbol error probability** of the channel ( $0 \leq p \leq 1$ ).

**Binary Symmetric Channel (BSC)**

A 2-symmetric channel is called a binary symmetric channel.



For a BSC:

1. If  $p = 0$ , the channel is *perfect*.
2. If  $p = 1/2$ , the channel is *useless*.
3. If  $1/2 < p \leq 1$ , then flipping all received bits converts the channel to a BSC with  $0 \leq p < 1/2$ .
4. Henceforth, we will assume that  $0 < p < 1/2$  for a BSC.

**Exercise:**

For a  $q$ -symmetric channel, show that one can take  $0 < p < \frac{q-1}{q}$  WLOG.

One can first consider the case  $q = 3$ .

**information rate**

The **information rate** (or rate)  $R$  of an  $[n, M]$ -code  $C$  over  $A$  is  $R = \log_q \frac{M}{n}$ .

If  $C$  encodes messages that are  $k$ -tuples over  $A$  (so  $M = |A^k| = q^k$ ), then  $R = \frac{k}{n}$ .

**Note:**

$0 \leq R \leq 1$ . Ideally,  $R$  should be close to 1.

**Example:**

The rate of the binary code  $C = \{00000, 11100, 00111, 10101\}$  is  $R = \frac{2}{5}$ .

**Hamming distance**

The **Hamming distance** (or distance) between two  $n$ -tuples over  $A$  is the number of coordinate positions in which they differ.

The Hamming distance (or distance) of an  $[n, M]$ -code  $C$  is  $d(C) = \min\{d(x, y) : x, y \in C, x \neq y\}$ .

**Example:**

The distance of  $C = \{00000, 11100, 00111, 10101\}$  is  $d(C) = 2$ .

**Theorem 1.1: properties of Hamming distance**

For all  $x, y, z \in A^n$ ,

1.  $d(x, y) \geq 0$ , with  $d(x, y) = 0$  iff  $x = y$ .
2.  $d(x, y) = d(y, x)$ .
3.  $d(x, y) + d(y, z) \geq d(x, z)$  ( $\triangle$  inequality).

## 1.2 Decoding Strategy

**Example:**

Let  $C = \{00000, 11100, 00111, 10101\}$ .  $C$  is a  $[5, 4]$ -code over  $\{0, 1\}$  (a binary code).

**Error Detection** If  $C$  is used for error detection only, the strategy is the following: A received word  $r \in A^n$  is accepted if and only if  $r \in C$ .

**Error Correction** Let  $C$  be an  $[n, M]$ -code over  $A$  with distance  $d$ . Suppose  $c \in C$  is transmitted, and  $r \in A^n$  is received. The (channel) decoder must decide one of the following:

- (i) No errors have occurred; *accept*  $r$ .
- (ii) Errors have occurred; *correct*<sup>1</sup> (decode)  $r$  to a codeword  $c \in C$ ?
- (iii) Errors have occurred; *no correction is possible*.

### Nearest Neighbour Decoding

- (i) Incomplete Maximum Likelihood Decoding (IMLD):

If there is a unique codeword  $c \in C$  such that  $d(r, c)$  is minimum, then correct  $r$  to  $c$ . If no such  $c$  exists, then report that errors have occurred, but correction is not possible (ask for retransmission, or disregard information).

- (ii) Complete Maximum Likelihood Decoding (CMLD):

Same as IMLD, except that if there are two or more  $c \in C$  for which  $d(r, c)$  is minimum, correct  $r$  to an arbitrary one of these.

<sup>1</sup>Error correction does not guarantee that the channel decoder always makes the correct decision. For example, 00000  $\xrightarrow{\text{transmit}}$  11100 which is accepted.



Is IMLD a reasonable strategy?

### Theorem 1.2

IMLD chooses the codeword  $c$  for which the conditional probability

$$P(r|c) = P(r \text{ is received} | c \text{ is sent})$$

is largest.

**Proof:**

Suppose  $c_1, c_2 \in C$  with  $d(c_1, r) = d_1$  and  $d(c_2, r) = d_2$ . Suppose  $d_1 > d_2$ .

Now

$$P(r|c_1) = (1-p)^{n-d_1} \left( \frac{p}{q-1} \right)^{d_1}$$

and

$$P(r|c_2) = (1-p)^{n-d_2} \left( \frac{p}{q-1} \right)^{d_2}$$

So,

$$\frac{P(r|c_1)}{P(r|c_2)} = (1-p)^{d_2-d_1} \left( \frac{p}{q-1} \right)^{d_1-d_2} = \left( \frac{p}{(1-p)(q-1)} \right)^{d_1-d_2}$$

Recall

$$\begin{aligned} p < \frac{q-1}{q} &\implies pq < q-1 \implies 0 < q-pq-1 \\ \implies p < p+q-pq-1 &\implies p < (1-p)(q-1) \implies \frac{p}{(1-p)(q-1)} < 1 \end{aligned}$$

Hence

$$\frac{P(r|c_1)}{P(r|c_2)} < 1$$

and so

$$P(r|c_1) < P(r|c_2)$$

and the result follows.  $\square$

## Minimum Error Probability Decoding (MED)

An *ideal strategy* would be to correct  $r$  to a codeword  $c \in C$  for which  $P(c|r) = P(r \text{ is received} | c \text{ is sent})$  is largest. This is MED.

**Example: (IMLD/CMLD) is not the same as MED**

Consider  $C = \{000, 111\}$ . Suppose  $P(c_1) = 0.1$  and  $P(c_2) = 0.9$ . Suppose  $p = \frac{1}{4}$  (for a BSC).

$\begin{matrix} \uparrow & \uparrow \\ c_1 & c_2 \end{matrix}$

Suppose  $r = 100$  is the received word. Then

$$P(c_1|r) = \frac{P(r|c_1) \cdot P(c_1)}{P(r)} = \frac{p(1-p)^2 \times 0.1}{P(r)} = \frac{9}{640} \cdot \frac{1}{P(r)}$$

$$P(c_2|r) = \frac{P(r|c_2) \cdot P(c_2)}{P(r)} = \frac{(1-p)p^2 \times 0.9}{P(r)} = \frac{27}{640} \cdot \frac{1}{P(r)}$$

So, MED decodes  $r$  to  $c_2$ . But IMLD decodes  $r$  to  $c_1$ .

## IMLD vs. MED

- IMLD maximizes  $P(r|c)$ . MED maximizes  $P(c|r)$ .

- (i) MED has the drawback that the decoding algorithm depends on the probability distribution of source messages.
- (ii) If all source messages are equally likely, then CMLD and MED are equivalent:

$$P(r|c_i) = P(c_i|r) \cdot P(c_i)/P(r) = P(c_i|r) \cdot \underbrace{\left[ \frac{1}{M \cdot P(r)} \right]}_{\text{does not depend on } c_i}$$

(iii) In practice IMLD (or CMLD) is used.

- In this course, we will use IMLD/CMLD.

## 1.3 Error Correcting & Detecting Capabilities of a Code

### Detection Only

*Strategy:* If  $r$  is received, then accept  $r$  if and only if  $r \in C$ .

#### e-error detecting code

A code  $C$  is an  **$e$ -error detecting code** if the decoder always makes the correct decision if  $e$  or fewer errors per codeword are introduced by the channel.

**Example:**

Consider  $C = \{000, 111\}$ .

$C$  is a 2-error detecting code.

$C$  is not a 3-error detecting code.

#### Theorem 1.3

A code  $C$  of distance  $d$  is a  $(d-1)$ -error detecting code (but is not a  $d$ -error detecting code).

**Proof:**

Suppose  $c \in C$  is sent.

If no errors occur, then  $c$  is received (and is accepted).

Suppose that # of errors is  $\geq 1$  and  $\leq d-1$ ; let  $r$  be the received word. Then  $1 \leq d(r, c) \leq d-1$ , so  $r \notin C$ . Thus  $r$  is rejected. This proves that it is  $(d-1)$ -error detecting code.

Since  $d(C) = d$ , there exist  $c_1, c_2 \in C$  with  $d(c_1, c_2) = d$ . If  $c_1$  is sent and  $c_2$  is received, then  $c_2$  is accepted; the  $d$  errors go undetected.  $\square$

### Correction

*Strategy:* IMLD/CMLD

#### e-error correcting code

A code  $C$  is an  **$e$ -error correcting code** if the decoder always makes the correct decision if  $e$  or fewer errors per codeword are introduced by the channel.

**Example:**

Consider  $C = \{000, 111\}$ .

$C$  is a 1-error correcting code.

$C$  is not a 2-error correcting code.

**Theorem 1.4**

A code  $C$  of distance  $d$  is an  $e$ -error correcting code, where  $e = \lfloor \frac{d-1}{2} \rfloor$ .

**Proof:**

Suppose that  $c \in C$  is sent, at most  $\frac{d-1}{2}$  errors are introduced, and  $r$  is received. Then  $d(r, c) \leq \frac{d-1}{2}$ .

On the other hand, if  $c_1$  is any other codeword, then

$$\begin{aligned} d(r, c_1) &\geq d(c, c_1) - d(r, c) && \triangle \text{ ineq} \\ &\geq d - \frac{d-1}{2} && \text{since } d(C) = d \\ &= \frac{d+1}{2} \\ &> \frac{d-1}{2} \geq d(r, c) \end{aligned}$$

Hence  $c$  is the unique codeword at minimum distance from  $r$ , so the decoder correctly concludes that  $c$  was sent.  $\square$

**Exercise:**

Suppose  $d(C) = d$ , and let  $e = \lfloor \frac{d-1}{2} \rfloor$ . Show that  $C$  is *not* an  $(e+1)$ -error correcting code.

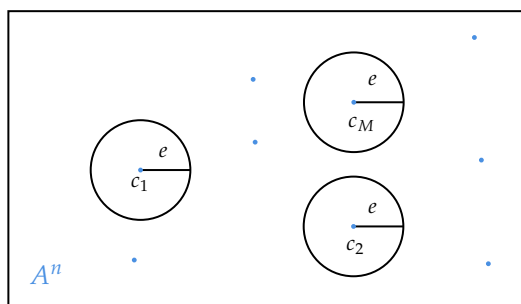
A natural question to ask is: given  $A, n, M, d$ , does there exist an  $[n, M]$ -code  $C$  over  $A$  of distance  $\geq d$ . This can be phrased as an equivalent sphere packing problem:

**Sphere packing**

Can we place  $M$  spheres of radius  $e = \lfloor \frac{d-1}{2} \rfloor$  in  $A^n$  so that no two spheres overlap?

$C = \{c_1, \dots, c_M\}$ ,  $e = \lfloor \frac{d-1}{2} \rfloor$ ,  $S_c$  = sphere of radius  $e$  centered at  $c$  = all words within distance  $e$  of  $c$ .

We proved: if  $c_1, c_2 \in C$ ,  $c_1 \neq c_2$ , then  $S_{c_1} \cap S_{c_2} = \emptyset$ .



Let  $n = 128, q = 2, M = 2^{64}$ . Does there exist a binary  $[n, M]$ -code with  $d \geq 22$ ? If so, can encoding and decoding be done efficiently?

We'll view  $\{0, 1\}^{128}$  as a vector space of dimension 128 over  $\mathbb{Z}_2$ . We'll choose  $C$  to be a 64-dimensional subspace of this vector space. We will construct such a code at the end of the course. The main tools used will be linear algebra (over finite fields) and abstract algebra (rings and fields).

# Introduction to Finite Fields

## 2.1 Definitions

### ring

A **(commutative) ring**  $(R, +, \cdot)$  consists of a set  $R$  and two operations  $+: R \times R \rightarrow R$  and  $\cdot: R \times R \rightarrow R$ , such that

1.  $a + (b + c) = (a + b) + c \quad \forall a, b, c \in R.$
2.  $a + b = b + a, \quad \forall a, b \in R.$
3.  $\exists 0 \in R$  such that  $a + 0 = a, \forall a \in R.$
4.  $\forall a \in R, \exists -a \in R$  such that  $a + (-a) = 0.$
5.  $a \cdot (b \cdot c) = (a \cdot b) \cdot c, \quad \forall a, b, c \in R.$
6.  $a \cdot b = b \cdot a, \quad \forall a, b \in R.$
7.  $\exists 1 \in R, 1 \neq 0$ , such that  $a \cdot 1 = a, \quad \forall a \in R.$
8.  $a \cdot (b + c) = a \cdot b + a \cdot c, \quad \forall a, b, c \in R.$

**Notation** We will denote  $(R, +, \cdot)$  by  $R$ .

**Example:**

$\mathbb{Q}, \mathbb{R}, \mathbb{C}, \mathbb{Z}$  are commutative rings.

### field

A **field**  $(F, +, \cdot)$  is a commutative ring with the additional property:

9.  $\forall a \in F, a \neq 0, \exists a^{-1} \in F$  such that  $a \cdot a^{-1} = 1.$

**Example:**

$\mathbb{Q}, \mathbb{R}, \mathbb{C}$  are fields.  $\mathbb{Z}$  is *not* a field.

**infinite/finite field**

A field  $(F, +, \cdot)$  is a **finite field** if  $F$  is a finite set; otherwise it is an **infinite field**. If  $F$  is a finite field, its **order** is  $|F|$ .

**Example:**

$\mathbb{Q}, \mathbb{R}, \mathbb{C}$  are infinite fields.

For which integers  $n \geq 2$  does there *exist* a finite field of order  $n$ ? How does one *construct* such a field, i.e., what are the field elements, and how are the field operations performed?

**The Integers Modulo  $n$** 

Let  $n \geq 2$ . Recall that  $\mathbb{Z}_n$  consists of the set of equivalence classes of integers modulo  $n$ ,  $\mathbb{Z}_n = \{[0], [1], [2], \dots, [n-1]\}$ , with addition and multiplication:  $[a] + [b] = [a + b]$ ,  $[a] \cdot [b] = [a \cdot b]$ .

More simply, we write  $\mathbb{Z}_n = \{0, 1, 2, \dots, n-1\}$ , and perform addition and multiplication modulo  $n$ .

**Example:**

$\mathbb{Z}_9 = \{0, 1, 2, \dots, 8\}$ . In  $\mathbb{Z}_9$ ,  $3 + 7 = 1$  and  $3 \cdot 7 = 3$ .

More precisely,  $3 + 7 \equiv 1 \pmod{9}$  and  $3 \cdot 7 \equiv 3 \pmod{9}$ .

$\mathbb{Z}_n$  is a commutative ring (i.e., axioms 1-8 in the definition are satisfied).

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