

15/1/2020

## Lecture 3

(MAP Decoding,  
ML Decoding  
Minimum distance  
of a block code)

### Announcements

- ① Tomo 10:30 am Tutorial (16th Jan)
- ② Assignment 2 (graded) - Released on 16th Jan, Due on 22nd Jan, Discussion will be on 23rd Jan

Assignment 2 will be on linear codes (Main material required to solve the assignment will be covered on Wednesday).

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Block Codes ( $n$  - Blocklength  
or length of the code).

$M$  - No. of Codewords in the code.

Each codeword is an  $n$ -tuple over the code alphabet  $F$ .

Decoder:  $D: F^n \rightarrow C$ .

$$P_e(\underline{c}) = \sum_{\underline{y}: \mathcal{D}(\underline{y}) \neq \underline{c}} \Pr[\underline{y} | \underline{c}]$$

Average probability of error (Code & Decoder have to be specified)

$P_e(\underline{c})$  averaged over all possible Codewords in the code

$$P_e = \sum_{\underline{c} \in \mathcal{C}} P_e(\underline{c}) R(\underline{c})$$

↳ Total probability Theorem

There is a decoder known as MAP Decoder.

MAP stands for Maximum A-posteriori decoder which is optimum decoder for a given code & channel.

optimum decoder means that it minimizes the average probability of error.

$$P_e = \sum_{\underline{c} \in \mathbb{C}} \sum_{\underline{y}: \mathcal{D}(\underline{y}) \neq \underline{c}} P_r(\underline{y} | \underline{c}) P(\underline{c}).$$

Interchanging  
the  
summations.

$$= \sum_{\underline{y}} \sum_{\substack{\underline{c} \in \mathbb{C}: \\ \mathcal{D}(\underline{y}) \neq \underline{c}}} P_r(\underline{y} | \underline{c}) P(\underline{c}).$$

$$= \sum_{\underline{y}} \sum_{\substack{\underline{c} \in \mathbb{C}: \\ \mathcal{D}(\underline{y}) \neq \underline{c}}} P(\underline{c} | \underline{y}) P(\underline{y}).$$

$$= \sum_{\underline{y}} P(\underline{y}) \left( \sum_{\substack{\underline{c} \in \mathbb{C}: \\ \mathcal{D}(\underline{y}) \neq \underline{c}}} P(\underline{c} | \underline{y}) \right)$$

$$\mathcal{D}: \mathbb{F}^n \rightarrow \mathbb{C}.$$

$$\mathcal{D}(\underline{y}) \rightarrow$$

↓  
Optimum decoder has to  
minimize the sum  $\neq \underline{y}$ .

What choice of  $\underline{c}$  minimizes the sum.

$$\left( \sum_{\substack{\underline{c} \in \mathbb{C}: \\ \mathcal{D}(\underline{y}) \neq \underline{c}}} P(\underline{c} | \underline{y}) \right)$$

$$\sum_{\substack{\underline{c} \in \mathbb{C}: \\ \mathcal{D}(\underline{y}) \neq \underline{c}}} P(\underline{c} | \underline{y}) = \sum_{\underline{c} \in \mathbb{C}} P(\underline{c} | \underline{y}) - P(\hat{\underline{c}} | \underline{y}).$$

$\xrightarrow{\text{Does not depend on } \mathcal{D}} \underline{c} \in \mathbb{C}$

The term on the left is minimized if the second term on the right is maximized

$$\hat{\underline{c}} = \arg \max_{\underline{c} \in \mathbb{C}} P_r(\underline{c} | \underline{y}) \quad \checkmark$$

MAP Decoding Rule.

$\hat{\underline{c}}$  is a function of  $\underline{y}$ .

$$\mathcal{D}: \mathbb{F}^n \rightarrow \mathbb{C}:$$

$$\underline{y} \rightarrow \hat{\underline{c}}:$$



### Maximum Likelihood Decoder (ML Decoder)

$$\arg \max_{\underline{c} \in \mathbb{C}} P_r(\underline{c} | \underline{y}) = \arg \max_{\underline{c} \in \mathbb{C}} \frac{P_r(\underline{y} | \underline{c}) P(\underline{c})}{P(\underline{y})}$$

If  $P(\underline{c}) = \binom{1/M}$  constant; which means that all codewords are equally likely.

$$\arg \max_{\underline{c} \in \mathbb{C}} P_r(\underline{c} | \underline{y}) = \arg \max_{\underline{c} \in \mathbb{C}} \underbrace{P_r(\underline{y} | \underline{c})}_{\text{Likelihood}}$$

$$\hat{c} = \arg \max_{c \in \mathbb{C}} P_r(\underline{y} | c).$$

↳ Maximum likelihood Decoder.

ML Decoder is same as MAP Decoder if codewords are equally likely.

→ ML Decoding is said to be optimum

Example of ML Decoding applied to the case of BSC with cross over probability  $p$ .

$$P_r(\underline{y} | c) = p^d \cdot (1-p)^{n-d}.$$

$d_H(\underline{y}, c) \triangleq$  Hamming distance between  $\underline{y}$  and  $c$ .  
 $= d.$

Ex -  $\underline{y} = 101110.$  →  $d_H(\underline{y}, c) = 3.$   
 $\underline{c} = 011010.$

→ Given a received vector  $\underline{y}$ .

$$\hat{c} = \arg \max_{c \in \mathbb{C}} P_r(\underline{y} | c).$$

$$= \arg \max_{c \in \mathbb{C}} p^d (1-p)^{n-d} = \arg \max_{c \in \mathbb{C}} (1-p)^n \left( \frac{p}{1-p} \right)^d$$

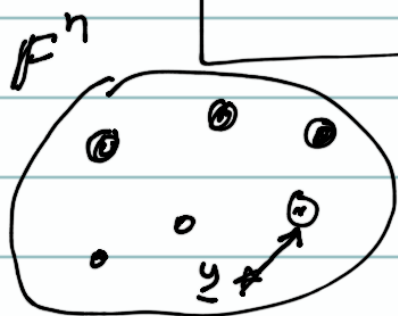


If  $\frac{p}{1-p} \leq 1$ , i.e.,  $p \leq \frac{1}{2}$ ;

then  $\hat{\underline{c}} = \arg \max_{\underline{c} \in \mathbb{C}} (1-p)^n \left( \frac{p}{1-p} \right)^d$ .

$d$  has to be minimized.

$$\hat{\underline{c}} = \arg \min_{\underline{c} \in \mathbb{C}} d_H(\underline{y}, \underline{c})$$



$$\underline{\mathbb{C}} \subseteq \mathbb{F}^n$$

→ ML decoding is equivalent to Minimum distance decoding

By distance, we mean Hamming distance.

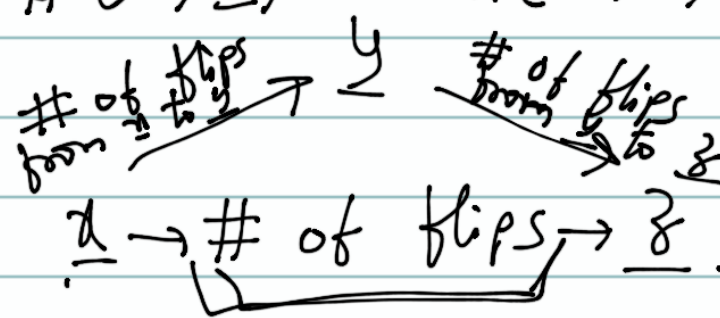
### Hamming Distance

Hamming distance between two vectors  $\underline{x}$  and  $\underline{y} \in \mathbb{F}^n$  is the no. of positions/coordinates

in which  $\underline{x}$  and  $\underline{y}$  differ.

$$\begin{array}{r} \underline{x} \quad 1 \ 0 \ 2 \ 1 \ 1 \ 0 \\ \underline{y} \quad 0 \ 0 \ 1 \ 0 \ 1 \ 0 \end{array} \quad d_H(\underline{x}, \underline{y}) = 3$$

## Properties:-

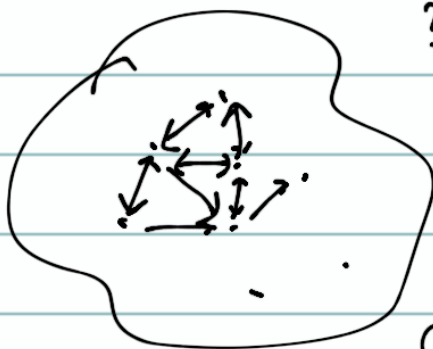
- ①  $d_H(\underline{x}, \underline{y}) \geq 0$ . (Positive)
- ②  $d_H(\underline{x}, \underline{y}) = d_H(\underline{y}, \underline{x})$ . (Symmetric).
- ③  $d_H(\underline{x}, \underline{z}) \leq d_H(\underline{x}, \underline{y}) + d_H(\underline{y}, \underline{z})$ .  
(Triangle Inequality).  


## Parameter of a Block Code

### Minimum Distance of a Block Code

Defn:-  $d(C) = d_{\min}(C)$

$$= \min_{\substack{\underline{x}, \underline{y} \in C \\ \underline{x} \neq \underline{y}}} d_H(\underline{x}, \underline{y}).$$



$(n, M)$  block Code

$(n, M, d)$  block Code.

$d$  = minimum distance of the block code.

## Examples of Codes

- ①  $n$ -fold repetition.  $\rightarrow (n, 2, n)$
- ② Simple parity check  $\rightarrow (n, 2^{n-1}, 2)$   
Code.
- ③  $(7, 4)$  Hamming Code.  $\rightarrow (7, 2^4, 3)$   
↑

$\rightarrow$  Minimum distance of a code determines the error detection capability & error correction capability.

## Error Detection

Consider a  $(n, M, d)$  block code  $\mathcal{C}$

Claim: There is a decoder  $\mathcal{D}$  for the code  $\mathcal{C}$  which can detect <sup>upto</sup>  $(d-1)$  errors.

Proof:

$$\mathcal{D}(\underline{y}) = \begin{cases} \underline{y} & \text{if } \underline{y} \in \mathcal{C}. \\ \text{Error} & \text{otherwise.} \end{cases}$$

$\underline{c} \rightarrow \underline{y}$   
s.t.  $d_H(\underline{y}, \underline{c}) \leq d-1$   $\uparrow$  If  $d_H(\underline{y}, \underline{c}) \neq 0$   
&  $d_H(\underline{y}, \underline{c}) \leq d-1$ ,  
then  $\underline{y} \notin \mathcal{C}$ .



## Error Correction

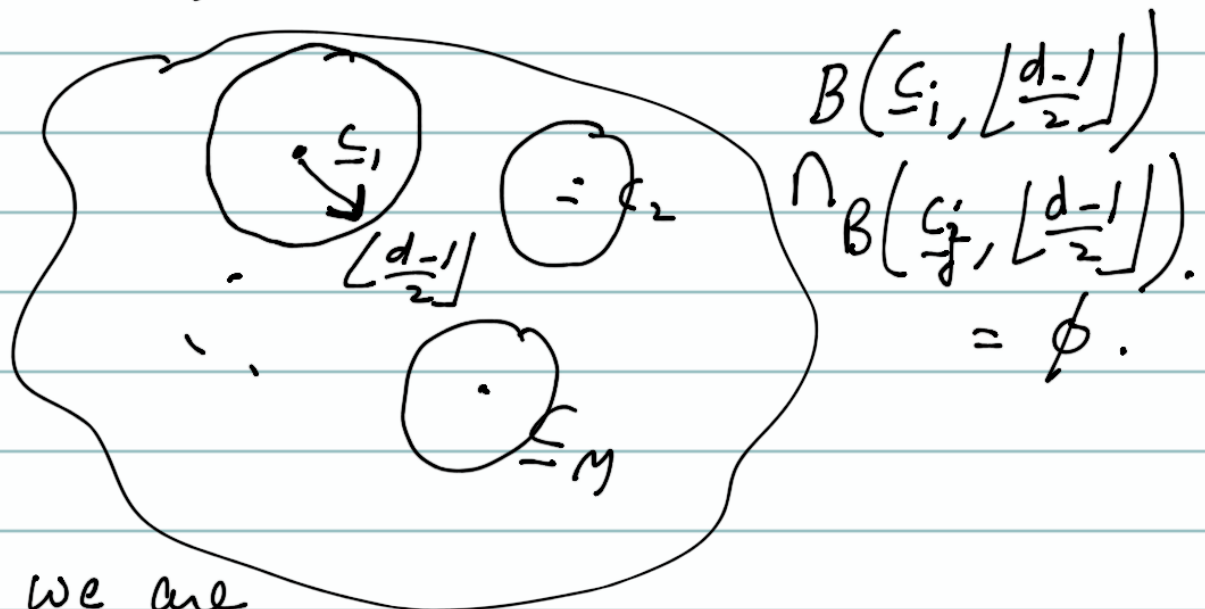
Consider a  $(n, M, d)$  block code  $\mathbb{C}$ .

claim: There is a decoder for  $\mathbb{C}$  that corrects up to  $\lfloor \frac{d-1}{2} \rfloor$  errors.

Proof: Decoder is minimum distance decoder.

$$D(\underline{y}) = \arg \min_{\underline{c} \in \mathbb{C}} d_H(\underline{y}, \underline{c}).$$

$\lfloor x \rfloor$  = largest integer less than or equal to  $x$ .



Ball that we are drawing around each codeword is a Hamming ball.

$$B(\underline{c}, t) = \{ \underline{y} \in \mathbb{F}^n : d_H(\underline{y}, \underline{c}) \leq t \}$$

$$d_H(\underline{c}_i, \underline{c}_j) \geq d$$

$$\underline{y} \in B(\underline{c}_i, \lfloor \frac{d-1}{2} \rfloor)$$

$$d_H(\underline{y}, \underline{c}_i) \leq \lfloor \frac{d-1}{2} \rfloor.$$

$$d \leq d_H(\underline{c}_i, \underline{c}_j) \leq d_H(\underline{y}, \underline{c}_i) + d_H(\underline{y}, \underline{c}_j).$$

$$d_H(\underline{y}, \underline{c}_j) \geq d - \lfloor \frac{d-1}{2} \rfloor.$$

$$> \lfloor \frac{d-1}{2} \rfloor$$

$$\underline{y} \notin B(\underline{c}_j, \lfloor \frac{d-1}{2} \rfloor).$$

All the Hamming balls are disjoint.