TSKS01 Digital Communication Lecture 10

Bounds on good codes, convolutional codes

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### Last two times – Binary Linear Codes [n, k, d]

A vector space expressed in a ba	asis	the nullspace of a matrix				
$\mathcal{C} = \{ \overline{m} G \; \forall \overline{m} \in \mathbb{F}_2^k \}$			$\mathcal{C} = \{ \bar{c} \in \mathbb{F}_2^n : H\bar{c}^T = \bar{0} \}$			
Generator matrix $(k \times n)$ , linearly independent rows.	$HG^{\mathrm{T}}$	= 0	Parity check matrix $((n - k) \times n)$ , linearly independent rows.			

Length: n, # columns in G or H

Dimension: k, # rows in G.

#### Minimum distance, d

Smallest Hamming distance between different codewords.

Smallest Hamming weight of non-zero codewords.

Smallest number of linearly dependent columns in *H*.





### **Outline of This Lecture**

- Block codes
  - Benefits of error control codes
  - Soft decoding
  - Bounds on good codes
- Convolutional codes
  - Introduction





## Error Probability: First comparison 1(2)

**Uncoded communication** over BSC with error probability p

$$k = n = 57, d = 1$$

 $P_e = \Pr\{\text{at least one error among } k \text{ bits}\} \\= 1 - \Pr\{\text{no errors among } k \text{ bits}\} \\= 1 - (1 - p)^k \approx kp$ 



Coded communication over the same BSC

Encoding: Hamming [63,57,3]

k = 57, n = 63, d = 3

 $P_e = \Pr\{\text{at least two errors among } n \text{ bits}\} \\= 1 - \Pr\{\text{zero or one errors among } n \text{ bits}\} \\= 1 - (1 - p)^n - np(1 - p)^{n-1} \approx n(n - 1)p^2/2$ 



### Error Probability: First comparison 2(2)



Uncoded:  $P_e = 1 - (1 - p)^k$ 

Coded:  $P_e = 1 - (1 - p)^n - np(1 - p)^{n-1}$ 



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### Error Probability: Second comparison 1(2)

Unfair comparison if the total signal energy is different!

#### Uncoded and coded communication with same bit energy

Assume binary modulation and the same  $E_b$  in both cases:

• Use signal energy  $(k/n)E_b$  for each codeword bit

Result: BSC with error probability 
$$q = Q\left(\sqrt{\frac{k}{n}}Q^{-1}(p)\right)$$
 Note:  
 $n \ge k \rightarrow q \ge p$ 

Uncoded:  $P_e = 1 - (1 - p)^k$ Hamming [63,57,3] code:  $P_e = 1 - (1 - q)^n - nq(1 - q)^{n-1}$ 





### Error Probability: Second comparison 2(2)



Uncoded:  $P_e = 1 - (1 - p)^k$ 

Coded:  $P_e = 1 - (1 - q)^n - nq(1 - q)^{n-1}$ 

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### Soft Detection

• Recall: ML decision rules

**ML decision rule:** Set  $\hat{a} = a_i$  if  $f_{\bar{X}|A}(\bar{x}|a_k)$  is maximized for k = i.

**Equivalent ML rule:** Set  $\hat{a} = a_i$  if  $d(\bar{x}, \bar{s}_k)$  is minimized for k = i.

- These rules give mechanisms to make decisions
  - How certain are we that are our decisions are correct?
  - Hard decision:  $\hat{a} = a_i$
  - **Soft decision**:  $\hat{a} = a_i$  and likelihood of right decision

### Soft Detection: Likelihood of Being Right



**ML rule**: Select  $\hat{a} = a_1$  if  $f_{\bar{X}|A}(\bar{x}|a_1) > f_{\bar{X}|A}(\bar{x}|a_0)$ 

- Uncertain if  $f_{\bar{X}|A}(\bar{x}|a_1) \approx f_{\bar{X}|A}(\bar{x}|a_0)$
- Very certain if  $f_{\bar{X}|A}(\bar{x}|a_1) \gg f_{\bar{X}|A}(\bar{x}|a_0)$

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### Soft Detection: Log-Likelihood Ratio

Measure of certainty:

Is 
$$\frac{f_{\bar{X}|A}(\bar{x}|a_1)}{f_{\bar{X}|A}(\bar{x}|a_0)} \approx 1$$
 or  $\frac{f_{\bar{X}|A}(\bar{x}|a_1)}{f_{\bar{X}|A}(\bar{x}|a_0)} \gg 1$ ?

### **Definition: Log-likelihood ratio (LLR)**

$$LLR_{a_1,a_0}(\bar{x}) = \log\left(\frac{f_{\bar{X}|A}(\bar{x}|a_1)}{f_{\bar{X}|A}(\bar{x}|a_0)}\right)$$

- $LLR_{a_1,a_0}(\bar{x}) \approx 0$ : Uncertain
- LLR<sub> $a_1,a_0$ </sub>( $\bar{x}$ )  $\gg$  0: Certain that  $\hat{a} = a_1$
- LLR<sub> $a_1,a_0$ </sub>( $\bar{x}$ )  $\ll$  0: Certain that  $\hat{a} = a_0$

Useful information to evaluate end result



### **Example: Log-Likelihood Ratio**







# Example: Soft and Hard Decoding of Hamming Code Over AWGN channel





### What is a Good Block Code?

There only exist [n, k, d] codes for some n, k, and d

Natural questions:

- Given *n* and *k*, what is the largest minimal distance *d*?
- Given *n* and *d*, how many bits *k* can be sent?
- Given k and d, what is the smallest possible n?

Hard to answer exactly!

Mathematical bounds: Shows what does not exist

Computer simulation: Shows what does exit





### The Hamming Bound:

Based on packing spheres in the binary Hamming space.

A linear [n, k, d] code:

Size of code:  $2^k$  codewords.

Size of vector space:  $2^n$  vectors.

Decoding sphere of radius  $\lfloor (d-1)/2 \rfloor$  around each codeword.

Size of a sphere:  $\sum_{i=0}^{\lfloor (d-1)/2 \rfloor} \binom{n}{i}$ Size of union of spheres:  $2^k \sum_{i=1}^{\lfloor (d-1)/2 \rfloor} \binom{n}{i} \le 2^n$ 

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Perfect codes: Hamming codes

Definition: Perfect codes achieves Hamming bound

(All received signals can be corrected)

X

X

Х

Disjoint spheres

Repetition codes with odd n

Х

Х

Х

Х





### The Singleton Bound

Based on shortening of codewords

A linear [n, k, d] code:

Remove the same d - 1 coefficients in each codeword

Result: A linear [n', k', d'] code with n' = n - d + 1, k' = k,  $d' \ge 1$ .

Size of resulting code:  $2^{k'} = 2^k$  codewords.

Size of resulting vector space:  $2^{n'} = 2^{n-d+1}$  vectors.

Distinct codewords:  $2^{n'} \ge 2^{k'} \Rightarrow 2^{n-d+1} \ge 2^k$ 

Distinct codewords

Result:  $n - d + 1 \ge k$ 

Singleton bound:  $n - k \ge d - 1$ (Shows which [n, k, d] might be possible)

Equality achieved by:

**Repetition codes** 





# The Maximum Value of the Minimum Distance of Binary Linear Block Codes





## Upper and Lower Bounds on the Maximum Value of the Minimum Distance of Binary Linear Block Codes

k	10	11	12	13	14	15	16	17	18	19	20
n											
30	11	10	9	8	8	8	7	6	6	6	5
31	12	11	10	9	8	8	8	7	6	6	6
32	12	12	10	10	8-9	8	8	8	6-7	6	6
33	12	12	11	10	9-10	8-9	8	8	7-8	6-7	6
34	12	12	12	10	10	9-10	8-9	8	8	7-8	6-7
35	12-13	12	12	11	10	10	9-10	8	8	8	7-8
36	13-14	12-13	12	12	11	10	10	8-9	8	8	8
37	14	13-14	12-13	12	12	10-11	10	9-10	8-9	8	8
38	14	14	13-14	12	12	11-12	10-11	10	9-10	8-9	8
39	15	14	14	12-13	12	12	11-12	10-11	10	9-10	8-9
40	16	14-15	14	12-14	12-13	12	12	11-12	10-11	10	9-10

Data fetched from http://www.win.tue.nl/~aeb/voorlincod.html on March 14, 2005. © 2005 Mikael Olofsson, mikael@isy.liu.se No difference Difference is 1 Difference is 2





Difference Between Upper and Lower Bounds for the Maximum Value of the Minimum Distance of Binary Linear Block Codes



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### Example: Rate with Hamming code



Potential drawbacks: Large delay? Complicated decoding?



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### Introduction: Convolutional Codes



- Encoder: FIR filter, convolution
- State diagram, trellis
- Decoding: Viterbi algorithm

Input: Semi-infinite sequence Output: Semi-infinite sequence In practice: Not infinitely long







### **Binary Sequences**

Consider:  $a = a_0, a_1, a_2, \dots$  for  $a_i \in \mathbb{F}_2$ 

D-transform: Write sequence using delay operator *D* 

• 
$$A(D) = a_0 + a_1 D + a_2 D^2 + \cdots$$

Filtering of binary sequences:

$$A(D) \longrightarrow P(D) \longrightarrow X(D)$$

• 
$$P(D) = p_0 + p_1 D + p_2 D^2 + \dots + p_m D^m$$

- Consider: X(D) = A(D)P(D)
- Time representation x = a \* p:

$$x_m = \sum_{k=0}^m a_k p_{m-k} = \sum_{k=0}^m a_{m-k} p_k$$





### Impulse Response



Impulse response:

- $a^{(0)} = 1,0, \dots$  gives outputs  $c^{(0)} = 1,0,1,0, \dots$  and  $c^{(1)} = 1,1,1,0, \dots$
- A(D) = 1 gives output  $C^{(0)}(D) = 1 + D^2$  and  $C^{(1)}(D) = 1 + D + D^2$
- Gather in a generator matrix:

$$G(D) = (1 + D^2, 1 + D + D^2)$$



### Generating a Convolutional Code

- Dimensions:
  - k inputs to encoder
  - n outputs from encoder

Coding rate: 
$$R = \frac{k}{n}$$

- Generator matrix G(D)
  - Dimension *k x n*
- Generating codewords
  - Input:  $A(D) = a_0 + a_1 D + a_2 D^2 + \cdots$
  - Output: C(D) = A(D)G(D)

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