Digital Communication Chapter 6 & 7

Ola Jetlund

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### Channel Coding Part I

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Shannon

Linear Block Codes

System model
Encoding
Decoding Linear boodes
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Linear Block Codes

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Codes Codes

- Evaluating the average probability of symbol error for different bandpass modulation schemes
- ► Comparing different modulation schemes based on their error performances.

# This Week (and next week)

- ► Channel Coding
- ► Linear Block Codes
- Convolutional Codes

### Channel Coding Part I

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## Digital Communication System

### Channel Coding Part I

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#### Introduction

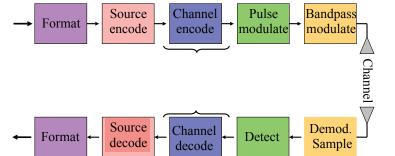
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For a DCS we can define the following goals:

- Maximizing the transmission bit rate
- Minimizing probability of bit error
- Minimizing the required power
- Minimizing required system bandwidth
- Maximizing system utilization
- Minimize system complexity

These are goals that must considered in the design phase.

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- Transforming signals to improve communication performance
- Waveform coding
- Structured sequences

### Error Control Techniques

- ARQ Automatic repeat request
- ▶ FEC Forward Error Correction
- ► Hybrid ARQ (ARQ + FEC)

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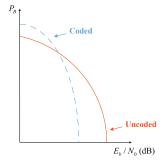
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Error performance vs. bandwidth

- Power vs. bandwidth
- Data rate vs. bandwidth
- Capacity vs. bandwidth BP

CODING GAIN: Reduction in  $E_b/N_0$  from using a code

$$G = \left(\frac{E_b}{N_0}\right)_u - \left(\frac{E_b}{N_0}\right)_c$$



- Use as few bits as possible to transmit information on a noisy channel
- Decode received information with as few errors as possible
- $\Rightarrow$  Utilize the channel capacity<sup>1</sup>:

$$C = W \log_2 \left( 1 + \frac{S}{N} \right)$$

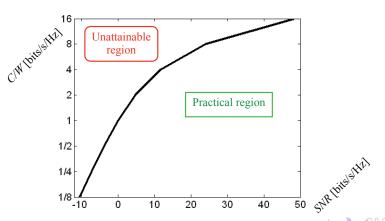
Here: capacity for an AWGN channel with:

- W Bandwidth [Hz]
- S Average received signal power
- N Average Noise power

### The Shannon theorem:

A limit on transmission data rate  $R_b$ 

- ▶ Transmission with  $R_b \le C$  is possible
  - with an arbitrary small error probability
- For R<sub>b</sub> > C, transmission cannot achieve an arbitrary small error probability.



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$$C = W \log_2 \left( 1 + \frac{S}{N} \right)$$

$$S = E_b C$$

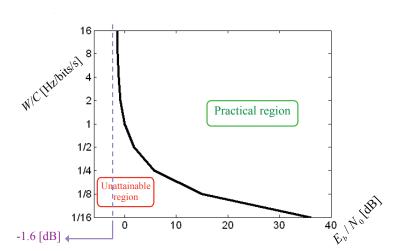
$$N = N_0 W$$

$$\Rightarrow \frac{C}{W} = \log_2 \left( 1 + \frac{E_b}{N_0} \frac{C}{W} \right)$$

As 
$$W \to \infty$$
 or  $\frac{C}{W} \to 0$ ,  $\frac{E_b}{N_0} \to \frac{1}{\log_2(e)} = 0.693 \approx -1.6$  [dB]

- No error free transmission (at any rate) for  $\frac{E_b}{N_0} < -1.6$  [dB]
- Capicity can be increased by increasing the bandwidth

### The Shannon limit ...



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# Linear Block Codes (n, k), I

Coding is a mapping from one space to another:



 $V_m$  is a vector space containing all  $2^k$  sequences of length m.

- ▶ A set  $C \subset V_n$  with cardinality  $2^k$  is called a linear block code if and only if it is a subspace of the vector space  $V_n$ .
  - Members of C are called codewords
  - The all-zero word is a codeword
  - any linear combination of a codeword is a codeword

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► The set {0,1} under the modulo-2 binary addition and multiplication forms a field.

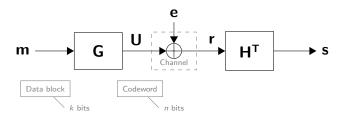
Addition	Multiplication	
0 ⊕ 0 = 0	0 ⊗ 0 = 0	
$0 \oplus 1 = 1$	$0 \otimes 1 = 0$	
$1 \oplus 0 = 1$	$1 \otimes 0 = 0$	
$1 \oplus 1 = 0$	$1\otimes 1 = 1$	

- ► AKA the Galois field: GF(2)
- ► Example V<sub>3</sub>

$$V_3 = \{(000), (001), (010), (011), (100), (101), (110), (111)\}$$

It has  $2^k = 2^3 = 8$  members (cardinality)

# Linear Block Codes (n, k), II



- m Message
- G Generator matrix
- **U** Codeword
- e Error introduced by channel

- r Received codeword
- H Parity check matrix
- s Syndrome (received data)

Note! We consider only binary sequences!

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► The (n, k) linear block code takes k data bit and produces a coded sequence of length n

- $\triangleright$  n-k parity bits
- Code rate:

$$R_c = \frac{k}{n}$$

► Coding:

$$U = m \cdot G$$

The generator matrix **G** is of size  $k \times n$  and it can be

- systematic, or
- nonsystematic.

$$\mathbf{G} = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 & 1 & 0 \end{bmatrix} = \begin{bmatrix} \mathbf{g_1} \\ \mathbf{g_2} \\ \mathbf{g_3} \\ \mathbf{g_4} \end{bmatrix}$$

Given  $\mathbf{m} = [1000]$ , then

$$\mathbf{U} = \mathbf{d} \cdot \mathbf{G} = [1100010]$$

Note that all rows in **G** are codewords.

Remember: A linear combination of two codewords are a codeword

Thus we can manipulate G as follows:

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$$\mathbf{g_5} = \mathbf{g_1} + \mathbf{g_2} = [1\,1\,0\,0\,0\,1\,0] + [0\,1\,0\,0\,1\,1\,1] = [1\,0\,0\,0\,1\,0\,1]$$

$$\mathbf{g_6} = \mathbf{g_3} + \mathbf{g_3} = [0\,0\,0\,1\,1\,1\,0] + [0\,0\,1\,1\,1\,1\,0] = [0\,0\,1\,0\,0\,0\,0]$$

Replace g<sub>1</sub> with g<sub>5</sub>, g<sub>4</sub> with g<sub>6</sub> to obtain new generator matrix:

Swap 3rd and 4th row in G' to obtain a new generator matrix:

$$\mathbf{G}^{\prime\prime} = \left[ \begin{array}{ccc|ccc|ccc|ccc|ccc|ccc|ccc|} 1 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 \end{array} \right]$$

Note that

$$\mathbf{G}'' = \left[ egin{array}{c|c} \mathbf{I}_4 & \mathbf{P} \end{array} 
ight]$$

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### For a systematic code

▶ the first *k* bits are the information bits:

$$\mathbf{G} = [\mathbf{I}_k \mid \mathbf{P}]$$

 $\mathbf{I}_k$  is a  $k \times k$  identity matrix  $\mathbf{P}$  is a  $k \times (n-k)$  matrix

► Thus,

$$\mathbf{U} = (u_1, u_2, \dots, u_n)$$
message bits parity bits
$$= (m_1, m_2, \dots, m_k, p_1, p_2, \dots, p_{n-k})$$

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For any linear block code with generator matrix **G** there exists a matrix **H** of size  $(n - k) \times n$  such that

$$\mathbf{G} \cdot \mathbf{H}^T = \mathbf{0}$$

- ▶ **H** is called the parity check matrix and its rows are linearly independent.
- ► For systematic linear block codes:

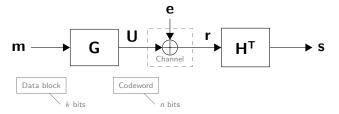
$$\mathbf{H} = [\mathbf{P}^T \mid \mathbf{I}_{n-k}]$$

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$$\mathbf{s} = \mathbf{r} \cdot \mathbf{H}^{\mathcal{T}} \qquad \mathbf{r} = \mathbf{U} + \mathbf{e} \qquad \mathbf{U} = \mathbf{m} \cdot \mathbf{G} \qquad \mathbf{G} \cdot \mathbf{H}^{\mathcal{T}} = \mathbf{0}$$

The syndrome **s**:

$$\mathbf{s} = \mathbf{r} \cdot \mathbf{H}^{T} = (\mathbf{U} + \mathbf{e}) \cdot \mathbf{H}^{T} = (\mathbf{m} \cdot \mathbf{G} + \mathbf{e}) \cdot \mathbf{H}^{T}$$
$$= \mathbf{m} \cdot \mathbf{G} \cdot \mathbf{H}^{T} + \mathbf{e} \cdot \mathbf{H}^{T} = \mathbf{m} \cdot \mathbf{0} + \mathbf{e} \cdot \mathbf{H}^{T}$$
$$= \mathbf{0} + \mathbf{e} \cdot \mathbf{H}^{T} = \mathbf{e} \cdot \mathbf{H}^{T}.$$



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 $s = r \cdot H^T = e \cdot H^T$ 

- 1. If s = 0 then r is a legal codeword and the decoded message  $\widehat{m}$  is found from  $\widehat{m}G = r$ ,
- 2. and if  $\mathbf{s} \neq \mathbf{0}$  then  $\mathbf{r}$  is a not legal codeword.
  - 2.1 Find an error vector  $\mathbf{e}'$  such that  $\mathbf{r} \mathbf{e}'$  is a legal codeword:

Find  $\mathbf{e}'$  such that  $(\mathbf{r} - \mathbf{e}') \cdot \mathbf{H}^T = \mathbf{0}$ 

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For a (n, k) linear block code there are  $2^k$  legal codewords:  $\mathbf{U}_i$ ,  $i \in {1, 2, ... } 2^k$ .

- ▶ Hamming weight:  $w(\mathbf{U}_i)$  = the number of non-zero elements in  $\mathbf{U}_i$ .
- ▶ Hamming distance:  $d(\mathbf{U}_i, \mathbf{U}_j) = w(\mathbf{U}_i \oplus \mathbf{U}_j)$
- Minimum distance:

$$d_{\min} = \min_{i \neq j} d(\mathbf{U}_i, \mathbf{U}_j) = \min_i w(\mathbf{U}_i)$$

A (n, k) linear block code can then

- ▶ Detect  $e = d_{\min} 1$  errors
- ▶ Correct  $t = \lfloor \frac{d_{\min} 1}{2} \rfloor$  errors

# Example: Systematic Hamming code (m)

- Number of bits in codeword:  $n = 2^m 1$
- Number of information bits:  $k = 2^m m 1$
- ▶ Number of parity bits: n k = m
- ► Code rate  $\frac{k}{n} = \frac{2^m m 1}{2^m 1} = 1 \frac{m}{2^m 1}$

For Hamming codes the columns in **H** represents all binary vectors of length  $2^{n-k}$  (except the all-zero codeword.

m	n	k
3	7	4
4	15	11
5	31	26

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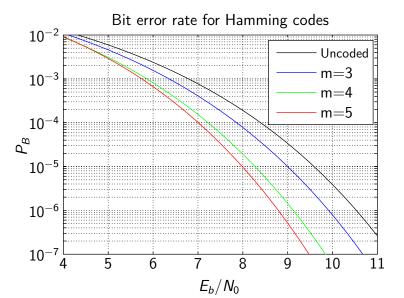
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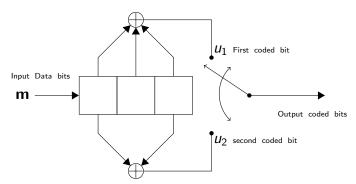
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Convolutional Codes

- Linear block codes
  - Rate  $R_c = \frac{k}{n}$
  - ► *n* is the length of the codewords
  - k is the length of the information sequence coded (or mapped) to one codeword
- Convolutional codes
  - Rate  $R_c = \frac{k}{n}$
  - n does not define a block or a codeword
  - k usually set to 1
  - ightharpoonup K constraint length counting the number of memory elements (which is K-1)

A convolutional code encodes the entire stream of data into a single codeword.

# Example: Rate $R_c = \frac{1}{2}$



Here, K = 3Let  $\mathbf{m} = (101)$  and find the output.

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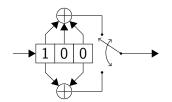
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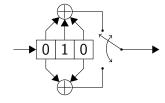
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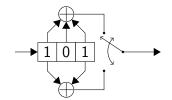
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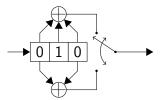






time 
$$t_1$$
:  $(u_1, u_2) = (1, 1)$  time  $t_2$ :  $(u_1, u_2) = (1, 0)$ 





time  $t_3$ :  $(u_1, u_2) = (0, 0)$  time  $t_4$ :  $(u_1, u_2) = (1, 0)$ 

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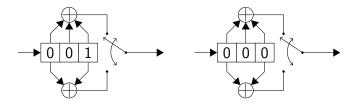
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time 
$$t_5$$
:  $(u_1, u_2) = (1, 1)$  time  $t_6$ :  $(u_1, u_2) = (0, 0)$ 

$$\mathbf{m} = (101) \longrightarrow \text{Encoder} \longrightarrow \mathbf{U} = (11\ 10\ 00\ 10\ 11)$$

### Description of convolutional codes

- A Tree diagrams (not very common)
- **B** State Diagram
  - Often used to find the exact error correcting properties of a code
- C The trellis diagram
  - Is often used to visualize the decoding process

Another larger will illustrate descriptions B and C.

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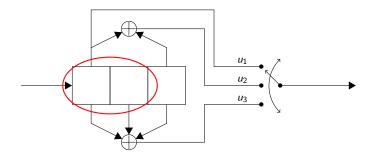
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# Example: $R_c = \frac{1}{3}$



The State of a convolutional code are the K-1 first bits in the encoder

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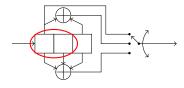
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### Inputs, states, and outputs



Input	Current State	Next State	Output	
0	0 0	00	000	"0"
1	0 0	10	111	"7"
0	01	00	011	"3"
1	01	10	100	" 4"
0	10	01	001	"1"
1	10	11	110	"6"
0	11	01	010	"2"
1	11	11	101	"5"

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State Diagram b 01 0/001 0/011 1/100 а C 0/000 0/010 00 10 1/111 1/110d 11

1/101

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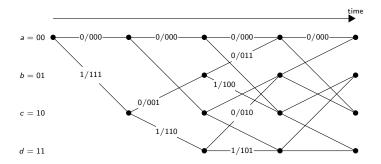
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## Trellis Diagram



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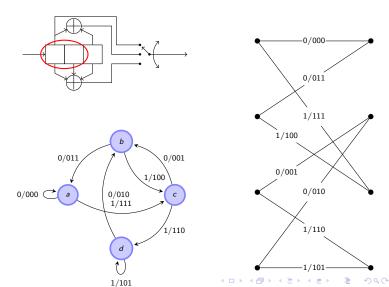
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# Convolutional codes (block -, state -, and trellis diagram



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