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# **16.682: Communication Systems Engineering**

## **Lecture 14: Channel Coding**

**April 5, 2001**

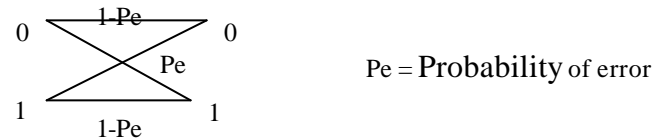
**Eytan Modiano**

# Channel Coding

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- When transmitting over a noisy channel, some of the bits are received with errors

## Example: Binary Symmetric Channel (BSC)



- **Q: How can these errors be removed?**
- **A: Coding: the addition of redundant bits that help us determine what was sent with greater accuracy**

# Example (Repetition code)

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Repeat each bit n times (n-odd)

Input	Code
0	000.....0
1	11.....1

## Decoder:

- If received sequence contains  $n/2$  or more 1's decode as a 1 and 0 otherwise
  - Max likelihood decoding

$$\begin{aligned} P(\text{error} \mid 1 \text{ sent}) &= P(\text{error} \mid 0 \text{ sent}) \\ &= P[\text{more than } n/2 \text{ bit errors occur}] \end{aligned}$$

$$= \sum_{i=\lceil n/2 \rceil}^n \binom{n}{i} P_e^i (1 - P_e)^{n-i}$$

# Repetition code, cont.

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- For  $P_e < 1/2$ ,  $P(\text{error})$  is decreasing in  $n$ 
  - $\exists$  for any  $\epsilon$ ,  $\exists n$  large enough so that  $P(\text{error}) < \epsilon$

## Code Rate: ratio of data bits to transmitted bits

- For the repetition code  $R = 1/n$
- To send one data bit, must transmit  $n$  channel bits “bandwidth expansion”
- In general, an  $(n,k)$  code uses  $n$  channel bits to transmit  $k$  data bits
  - Code rate  $R = k / n$
- Goal: for a desired error probability,  $\epsilon$ , find the highest rate code that can achieve  $p(\text{error}) < \epsilon$

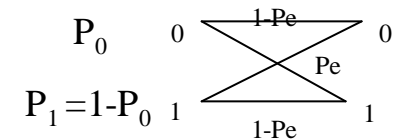
# Channel Capacity

- The capacity of a discrete memoryless channel is given by,

$$C = \max_{p(x)} I(X;Y)$$



## Example: Binary Symmetric Channel (BSC)



$$I(X;Y) = H(Y) - H(Y|X) = H(X) - H(X|Y)$$

$$H(X|Y) = H(X|Y=0) \cdot P(Y=0) + H(X|Y=1) \cdot P(Y=1)$$

$$H(X|Y=0) = H(X|Y=1) = P_e \log(1/P_e) + (1-P_e) \log(1/1-P_e) = H_b(P_e)$$

$$H(X|Y) = H_b(P_e) \Rightarrow I(X;Y) = H(X) - H_b(P_e)$$

$$H(X) = P_0 \log(1/P_0) + (1-P_0) \log(1/1-P_0) = H_b(p_0)$$

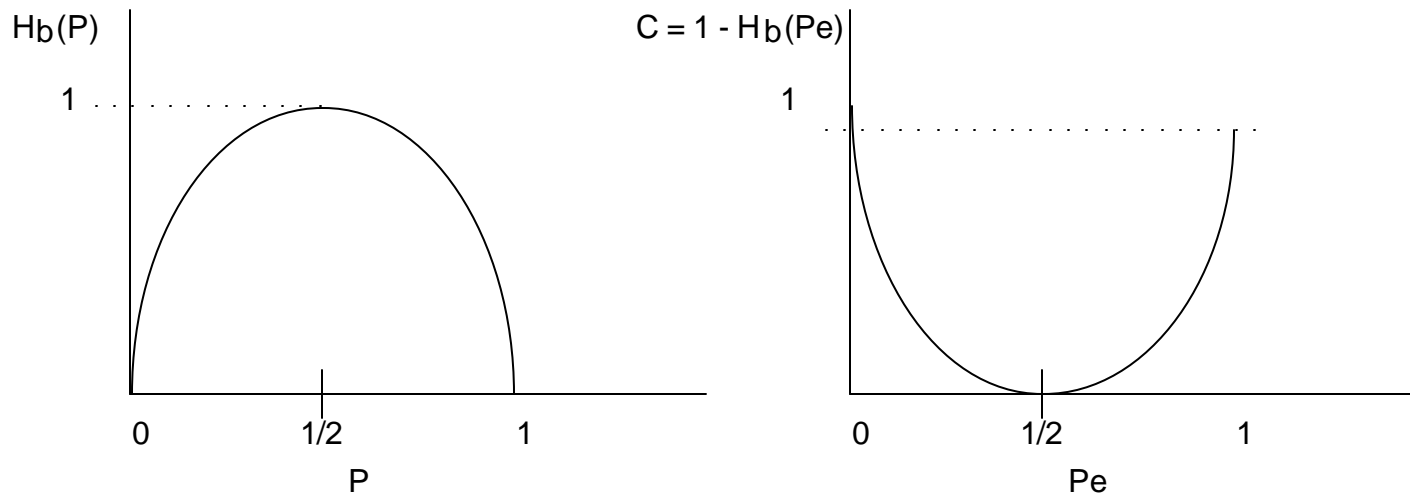
$$\Rightarrow I(X;Y) = H_b(P_0) - H_b(P_e)$$

# Capacity of BSC

$$I(X;Y) = H_b(P_0) - H_b(P_e)$$

- $H_b(P) = P \log(1/P) + (1-P) \log(1/1-P)$ 
  - $H_b(P) \leq 1$  with equality if  $P=1/2$

$$C = \max_{P_0} \{I(X;Y) = H_b(P_0) - H_b(P_e)\} = 1 - H_b(P_e)$$



**$C = 0$  when  $P_e = 1/2$  and  $C = 1$  when  $P_e = 0$  or  $P_e = 1$**

# Channel Coding Theorem (Claude Shannon)

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**Theorem: For all  $R < C$  and  $\epsilon > 0$ ; there exists a code of rate  $R$  whose error probability  $< \epsilon$**

- $\epsilon$  can be arbitrarily small
- Proof uses large block size  $n$   
as  $n \rightarrow \infty$  capacity is achieved

- **In practice codes that achieve capacity are difficult to find**

- The goal is to find a code that comes as close as possible to achieving capacity

- **Converse of Coding Theorem:**

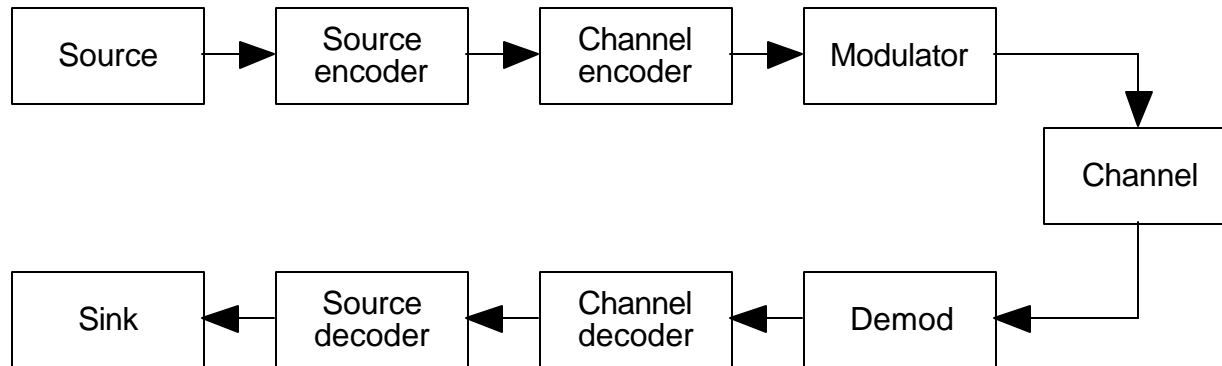
- For all codes of rate  $R > C$ ,  $\exists \epsilon_0 > 0$ , such that the probability of error is always greater than  $\epsilon_0$

For code rates greater than capacity, the probability of error is bounded away from 0

# Channel Coding

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- **Block diagram**





# Approaches to coding

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- **Block Codes**

- Data is broken up into blocks of equal length
- Each block is “mapped” onto a larger block

\_\_\_ **Example: (6,3) code,  $n = 6$ ,  $k = 3$ ,  $R = 1/2$**

**000 ® 000000**

**100 ® 100101**

**001 ® 001011**

**101 ® 101110**

**010 ® 010111**

**110 ® 110010**

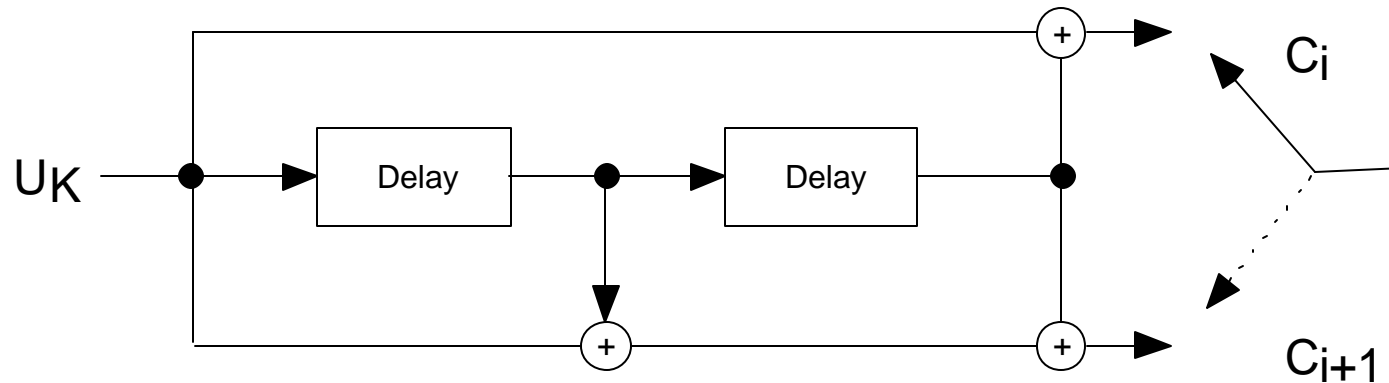
**011 ® 011100**

**111 ® 111001**

- **An (n,k) binary block code is a collection of  $2^k$  binary n-tuples ( $n > k$ )**
  - $n$  = block length
  - $k$  = number of data bits
  - $n-k$  = number of checked bits
  - $R = k / n$  = code rate

# Approaches to coding

- **Convolutional Codes**
  - The output is provided by looking at a sliding window of input



$$C_{(2K)} = U_{(2K)} \oplus U_{(2K-2)}, \quad C_{(2K+1)} = U_{(2K+1)} \oplus U_{(2K)} \oplus U_{(2K-1)}$$

$\oplus$  mod(2) addition ( $1+1=0$ )

# Block Codes

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- A block code is systematic if every codeword can be broken into a data part and a redundant part
  - Previous (6,3) code was systematic

## Definitions:

- Given  $X \in \{0,1\}^n$ , the Hamming Weight of  $X$  is the number of 1's in  $X$
- Given  $X, Y$  in  $\{0,1\}^n$ , the Hamming Distance between  $X$  &  $Y$  is the number of places in which they differ,

$$d_H(X, Y) = \sum_{i=1}^n X_i \oplus Y_i = \text{Weight}(X + Y)$$

$$X + Y = [x_1 \oplus y_1, x_2 \oplus y_2, \dots, x_n \oplus y_n]$$

- The minimum distance of a code is the Hamming Distance between the two closest codewords:

$$d_{\min} = \min_{C_1, C_2 \in C} \{d_H(C_1, C_2)\}$$

# Decoding

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- $r$  may not equal to  $u$  due to transmission errors
- Given  $r$  how do we know which codeword was sent?

## Maximum likelihood Decoding:

Map the received  $n$ -tuple  $r$  into the codeword  $C$  that maximizes,  
 $P \{ r \mid C \text{ was transmitted} \}$

## Minimum Distance Decoding (nearest neighbor)

Map  $r$  to the codeword  $C$  such that the hamming distance between  $r$  and  $C$  is minimized (i.e.,  $\min d_H(r, C)$ )

⊢ For most channels Min Distance Decoding is the same as Max likelihood decoding

# Linear Block Codes

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- A  $(n,k)$  linear block code (LBC) is defined by  $2^k$  codewords of length  $n$

$$C = \{ C_1, \dots, C_m \}$$

- A  $(n,k)$  LBC is a  $K$ -dimensional subspace of  $\{0,1\}^n$ 
  - $(0 \dots 0)$  is always a codeword
  - If  $C_1, C_2 \in C$ ,  $C_1 + C_2 \in C$
- **Theorem:** For a LBC the minimum distance is equal to the minimum weight ( $W_{\min}$ ) of the code

$$W_{\min} = \min_{(\text{over all } C_i)} \text{Weight}(C_i)$$

Proof: Suppose  $d_{\min} = d_H(C_i, C_j)$ , where  $C_1, C_2 \in C$

$$d_H(C_i, C_j) = \text{Weight}(C_i + C_j),$$

but since  $C$  is a LBC then  $C_i + C_j$  is also a codeword

# Systematic codes

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**Theorem:** Any  $(n,k)$  LBC can be represented in Systematic form

where:  $\text{data} = x_1 \dots x_k$ ,  $\text{codeword} = x_1 \dots x_k c_{k+1} \dots x_n$

- Hence we will restrict our discussion to systematic codes only
- The codewords corresponding to the information sequences:  $e_1 = (1, 0, \dots, 0)$ ,  $e_2 = (0, 1, 0, \dots, 0)$ ,  $e_k = (0, 0, \dots, 1)$  for a basis for the code
  - Clearly, they are linearly independent
  - $k$  linearly independent  $n$ -tuples completely define the  $k$  dimensional subspace that forms the code

Information sequence

$$e_1 = (1, 0, \dots, 0)$$

$$e_2 = (0, 1, 0, \dots, 0)$$

$$e_k = (0, 0, \dots, 1)$$

Codeword

$$g_1 = (1, 0, \dots, 0, g_{(1,k+1)} \dots g_{(1,n)})$$

$$g_2 = (0, 1, \dots, 0, g_{(2,k+1)} \dots g_{(2,n)})$$

$$g_k = (0, 0, \dots, 1, g_{(k,k+1)} \dots g_{(k,n)})$$

•  $g_1, g_2, \dots, g_k$  form a basis for the code

# The Generator Matrix

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$$G = \begin{bmatrix} g_1 \\ g_2 \\ \vdots \\ g_k \end{bmatrix} = \begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1n} \\ g_{21} & & & g_{2n} \\ \vdots & & & \\ g_{k1} & & & g_{kn} \end{bmatrix}$$

- For input sequence  $\mathbf{x} = (x_1, \dots, x_k)$ :  $\mathbf{C}_x = \mathbf{xG}$ 
  - Every codeword is a linear combination of the rows of  $G$
  - The codeword corresponding to every input sequence can be derived from  $G$
  - Since any input can be represented as a linear combination of the basis  $(\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_k)$ , every corresponding codeword can be represented as a linear combination of the corresponding rows of  $G$
- Note:  $x_1 \leftrightarrow \mathbf{C}_1, x_2 \leftrightarrow \mathbf{C}_2 \Rightarrow x_1 + x_2 \leftrightarrow \mathbf{C}_1 + \mathbf{C}_2$





# The parity check matrix

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$$H = \left[ \begin{array}{c|c} P^T & I_{(n-K)} \end{array} \right]$$

$I_{(n-K)}$  =  $(n - K) \times (n - K)$  identity matrix

Example:

$$H = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 1 & 1 & 0 & 0 & 1 \end{bmatrix}$$

Now, if  $c_i$  is a codeword of  $C$  then,  $c_i H^T = \vec{0}$

- “ $C$  is in the null space of  $H$ ”
- Any codeword in  $C$  is orthogonal to the rows of  $H$

# Decoding

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- $v =$  transmitted codeword  $= v_1 \dots v_n$
- $r =$  received codeword  $= r_1 \dots r_n$
- $e =$  error pattern  $= e_1 \dots e_n$
- $r = v + e$
- $S = rH^T =$  Syndrome of  $r$   
 $= (v+e)H^T = vH^T + eH^T = eH^T$
- $S$  is equal to '0' if and only if  $e \in C$ 
  - I.e., error pattern is a codeword
- $S \neq 0 \Rightarrow$  error detected
- $S = 0 \Rightarrow$  no errors detected (they may have occurred and not detected)
- Suppose  $S \neq 0$ , how can we know what was the actual transmitted codeword?

# Syndrome decoding

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- Many error patterns may have created the same syndrome

For error pattern  $e_0 \Rightarrow S_0 = e_0 H^T$

Consider error pattern  $e_0 + c_i$  ( $c_i \hat{I} C$ )

$$S'_0 = (e_0 + c_i)H^T = e_0 H^T + c_i H^T = e_0 H^T = S_0$$

- So, for a given error pattern,  $e_0$ , all other error patterns that can be expressed as  $e_0 + c_i$  for some  $c_i \hat{I} C$  are also error patterns with the same syndrome
- For a given syndrome, we can not tell which error pattern actually occurred, but the most likely is the one with minimum weight
  - Minimum distance decoding
- For a given syndrome, find the error pattern of minimum weight ( $e_{\min}$ ) that gives this syndrome and decode:  $r' = r + e_{\min}$

# Standard Array

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$M = 2^K$	$C_1$	$C_2$	$\dots$	$C_M$	<i>Syndrome</i>
	$e_1$	$e_1 + C_2$		$e_1 + C_M$	$S_1$
	$\vdots$	$e_2 + C_2$		$e_2 + C_M$	$S_2$
	$e_{2^{(n-K)}-1}$				$S_{2^{(n-K)}-1}$

- **Row 1 consists of all M codewords**
- **Row 2  $e_1$  = min weight n-tuple not in the array**
  - I.e., the minimum weight error pattern
- **Row i,  $e_i$  = min weight n-tuple not in the array**
  
- **All elements of any row have the same syndrome**
  - Elements of a row are called “co-sets”
- **The first element of each row is the minimum weight error pattern with that syndrome**
  - Called “co-set leader”

# Decoding algorithm

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- Receive vector  $r$ 
  - 1) Find  $S = rH^T =$  syndrome of  $r$
  - 2) Find the co-set leader  $e$ , corresponding to  $S$
  - 3) Decode:  $C = r+e$
- “Minimum distance decoding”
  - Decode into the codeword that is closest to the received sequence

# Example (syndrome decoding)

- Simple (4,2) code

$$G = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix}$$

<u>Data</u>	<u>codeword</u>
00	0000
01	0101
10	1010
11	1111

$$H = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} \quad H^T = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

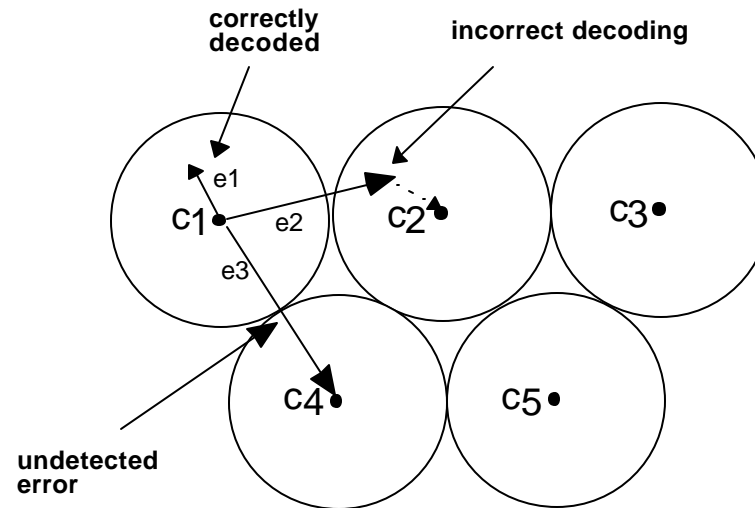
Standard array	0000	0101	1010	1111	Syndrome
	1000	1101	0010	0111	10
	0100	0001	1110	1011	01
	1100	1001	0110	0011	11

Suppose 0111 is received,  $S = 10$ , co-set leader = 1000

Decode:  $C = 0111 + 1000 = 1111$

# Minimum distance decoding

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- **Minimum distance decoding maps a received sequence onto the nearest codeword**
- **If an error pattern maps the sent codeword onto another valid codeword, that error will be undetected (e.g., e3)**
  - Any error pattern that is equal to a codeword will result in undetected errors
- **If an error pattern maps the sent sequence onto the sphere of another codeword, it will be incorrectly decoded (e.g., e2)**

# Performance of Block Codes

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- **Error detection: Compute syndrome,  $S \neq 0 \Rightarrow$  error detected**
  - Request retransmission
  - Used in packet networks
- **A linear block code will detect all error patterns that are not codewords**
- **Error correction: Syndrome decoding**
  - All error patterns of weight  $< d_{\min}/2$  will be correctly decoded
  - This is why it is important to design codes with large minimum distance ( $d_{\min}$ )
  - The larger the minimum distance the smaller the probability of incorrect decoding



# Hamming Codes

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- **Linear block code capable of correcting single errors**
  - $n = 2^m - 1$ ,  $k = 2^m - 1 - m$   
(e.g., (3,1), (7,4), (15,11)...)
  - $R = 1 - m/(2^m - 1) \Rightarrow$  very high rate
  - $d_{\min} = 3 \Rightarrow$  single error correction
- **Construction of Hamming codes**
  - Parity check matrix (H) consists of all non-zero binary m-tuples

**Example: (7,4) hamming code (m=3)**

$$H = \begin{bmatrix} 1 & 0 & 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 \end{bmatrix}, \quad G = \begin{bmatrix} 1 & 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix}$$