

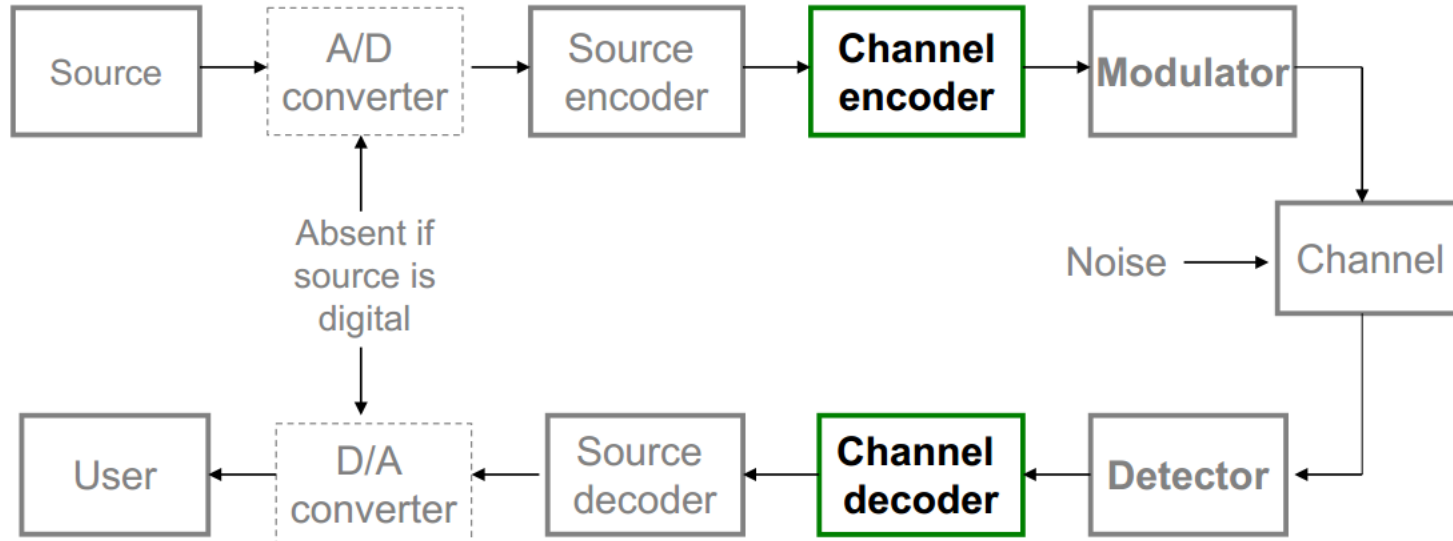


Outline

- Introduction
- Signal, random variable, random process and spectra
- Analog modulation
- Analog to digital conversion
- Digital transmission through baseband channels
- Signal space representation
- Optimal receivers
- Digital modulation techniques
- **Channel coding**
- Synchronization
- Information theory



Channel coding



- Linear block code
- Convolutional code

Chapter 13.1-13.3



Channel coding

- Information theory and channel coding
 - Shannon's **noisy channel coding theorem** tells us that adding controlled redundancy allows transmission at arbitrarily low bit error rate (BER) as long as $R \leq C$
 - **Error control coding (ECC)** uses this controlled redundancy to **detect** and **correct** errors
 - ECC depends on the system requirements and the nature of the channel
 - The key in ECC is to find a way to **add redundancy** to the channel so that the receiver can fully utilize that redundancy to detect and correct the errors, and to reduce the required transmit power (**coding gain**)



Channel coding

- Information theory and channel coding
 - Consider for example the case that we want to transmit data over a telephone link using a modem under the conditions that link bandwidth = 3 kHz and the modem can operate up to the speed of 3600 bits/sec at an error probability $P_e = 8 \times 10^{-4}$.
 - **Target:** transmit the data the rate of 1200 bits/sec at maximum output SNR = 13 dB with a probability of error 1×10^{-4}



Channel coding

- Information theory and channel coding
 - Shannon theorem tells us that channel capacity is

$$C = B \log_2 \left(1 + \frac{S}{N} \right) = 13,000 \text{ bits/sec}$$

since $B=3000$, $S/N=13 \text{ dB}=20$

- Thus, by Shannon's theorem, we can transmit the data with an arbitrarily small error probability
- Note that without coding $P_e = 8 \times 10^{-4}$, the target P_e is not met.



Channel coding

- Information theory and channel coding
 - Consider a simple code design with **repetition code**.
 - Every bit is transmitted 3 times, e.g., when $b_k = "0"$ or $"1"$, transmitted codewords are $"000"$ or $"111"$
 - Based on the received codewords, the decoder attempts to extract the transmitted bits using majority-logic decoding scheme
 - Obviously, the transmitted bits will be recovered correctly as long as no more than one of the bits in the codewords is affected by noise

Tx bits b_k	0	0	0	0	1	1	1	1
Codewords	000	000	000	000	111	111	111	111
Rx bits	000	001	010	100	011	101	110	111
\hat{b}_k	0	0	0	0	1	1	1	1



Channel coding

- Information theory and channel coding
 - With this simple error control coding, the probability of error is

$$P_e = P(b_k \neq \hat{b}_k)$$

$$= P(2 \text{ or more bits in codeword are in error})$$

$$= \binom{3}{2} q_c^2 (1 - q_c) + \binom{3}{3} q_c^3$$

$$= 3q_c^2 - 2q_c^3$$

$$= 0.0192 \times 10^{-4}$$

$$\leq \text{Required } P_e \text{ of } 10^{-4}$$



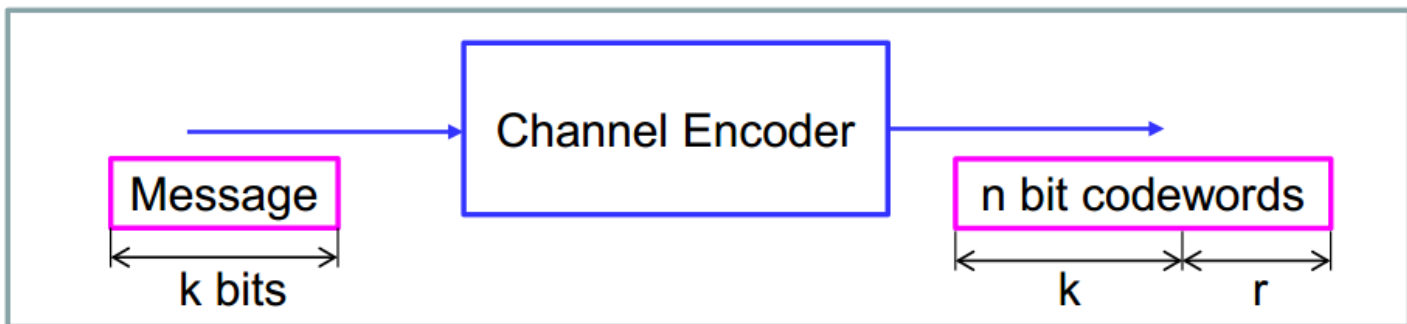
Channel coding

- From the above example, we can see the importance of coding techniques.
- Coding techniques are classified as either **block codes** or **convolutional codes**, depending on the presence or absence of **memory**
- A block code has no memory
 - **Information sequence** is broken into **blocks** of length **k**
 - Each block of **k** inf. bits is encoded into a block of **n** **coded bits**
 - No memory from one block to another block
- A convolutional code has memory
 - A **shift register** of length **k_0L** is used
 - **Inf. bits** enter the shift register **k_0** bits at a time and **no coded bits** are generated
 - These **no** bits depend not only on the **recent k_0 bits**, but also on the **$k_0(L-1)$ previous bits**



Linear block codes

- Block codes
 - An (n,k) block code is a collection of $M=2^k$ codewords of length n
 - Each codeword has a block of k inf. bits followed by a group of $r=n-k$ check bits that are derived from the k inf. bits in the block preceding the check bits



- The code is said to be linear if any linear combination of 2 codewords is also a codeword, i.e., if c_i and c_j are codewords, then $c_i + c_j$ is also a codeword (addition is module-2)



Linear Block codes

- **Code rate** (rate efficiency) = k/n
- Matrix description:
 - Codeword $\mathbf{c} = (c_1, c_2, \dots, c_n)$
 - Message bits $\mathbf{m} = (m_1, m_2, \dots, m_k)$
- Each block code can be generated using a **Generator matrix \mathbf{G}** (dim: $k \times n$)
- Given \mathbf{G} , then

$\mathbf{c} = \mathbf{m}\mathbf{G}$

Codeword message



Linear Block codes

- Generator matrix \mathbf{G}

$$\mathbf{G} = [\mathbf{I}_k | \mathbf{P}]_{k \times n}$$

$$= \left[\begin{array}{cccc|cccc} 1 & 0 & \cdots & 0 & p_{11} & p_{12} & \cdots & p_{1,n-k} \\ 0 & 1 & & 0 & p_{21} & p_{22} & & p_{2,n-k} \\ \vdots & & \ddots & \vdots & \vdots & & \ddots & \vdots \\ 0 & 0 & \cdots & 1 & p_{k,1} & p_{k,2} & \cdots & p_{k,n-k} \end{array} \right]$$

- \mathbf{I}_k is an identity matrix of order k
- \mathbf{P} is a matrix of order $k \times (n-k)$, which is selected so that the code will have certain desired properties



Linear Block codes

- Generator matrix \mathbf{G}
 - The form of \mathbf{G} implies that the 1st k components of any codeword are precisely the information symbols
 - This form of linear encoding is called **systematic encoding**
 - Systematic-form codes allow easy implementation and quick look-up features for decoding
 - For linear codes, any code is equivalent to a code in systematic form (given the same performance). Thus, we can restrict our study to only systematic codes



Linear Block codes

- Example

- Hamming code is a family of (n,k) linear block codes that have the following parameters

1. Codeword length $n = 2^m - 1, m \geq 3$
2. # of message bits $k = 2^m - m - 1$
3. # of parity check bits $n - k = m$
4. Capable of providing single-error correction capability with $d_{\min} = 3$

- $(7,4)$ Hamming code with generator matrix

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

Find all codewords



Linear Block codes

- Example
 - (7,4) Hamming code

Message				codeword							
0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	1	0	0	0	1	1	0	1	1
0	0	1	0	0	0	1	0	1	1	1	1
0	0	1	1	0	0	1	1	0	1	1	0
0	1	0	0	0	1	0	0	0	1	1	1
0	1	0	1	0	1	0	1	1	1	1	0
0	1	1	0	0	1	1	0	1	0	0	0
0	1	1	1	0	1	1	1	0	0	0	1
1	0	0	0	1	0	0	0	1	1	1	0
1	0	0	1	1	0	0	1	0	1	1	1
1	0	1	0	1	0	1	0	0	0	0	1
1	0	1	1	1	0	1	1	1	0	0	0
1	1	0	0	1	1	0	0	1	0	0	1
1	1	0	1	1	1	0	1	0	0	0	0
1	1	1	0	1	1	1	0	0	1	0	0
1	1	1	1	1	1	1	1	1	1	1	1



Linear Block codes

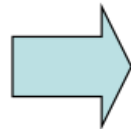
- Parity check matrix

- For each \mathbf{G} , it is possible to find a corresponding parity check matrix \mathbf{H}

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

- \mathbf{H} can be used to verify if a codeword \mathbf{C} is generated by \mathbf{G}

- Let \mathbf{C} be a codeword generated by $\mathbf{G} = [\mathbf{I}_k \mid \mathbf{P}]_{k \times n}$



$$\mathbf{cH}^T = \mathbf{mGH}^T = 0$$

- Think about the parity check matrix of (7,4) Hamming code



Linear Block codes

- Error syndrome

- Received codeword $\mathbf{r}=\mathbf{c}+\mathbf{e}$, where \mathbf{e} =Error vector or Error pattern and it is **1 in every position where data word is in error**

- Example

$$\mathbf{c} = [1 \ 0 \ 1 \ 0]$$

$$\mathbf{r} = [1 \ 1 \ 0 \ 0]$$

$$\mathbf{e} = [0 \ 1 \ 1 \ 0]$$

- Error syndrome: $\mathbf{s} \triangleq \mathbf{rH}^T$



Linear Block codes

- Error syndrome

- Note that

$$\begin{aligned} \mathbf{s} &= \mathbf{rH}^T = (\mathbf{c} + \mathbf{e})\mathbf{H}^T \\ &= \mathbf{cH}^T + \mathbf{eH}^T \\ &= \mathbf{eH}^T \end{aligned}$$

- If $\mathbf{s}=\mathbf{0}$, then $\mathbf{r} = \mathbf{c}$ and m is the 1st k bits of \mathbf{r}

- If $\mathbf{s}\neq\mathbf{0}$, and \mathbf{s} is the j th row of \mathbf{H}^T , then 1 error in j th position of \mathbf{r}



Linear Block codes

- Error syndrome

- Consider the (7,4) Hamming code for example

$$\mathbf{H}^T = [\mathbf{P}^T | \mathbf{I}_{n-k}]^T = \begin{bmatrix} \mathbf{P} \\ \mathbf{I}_{n-k} \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 0 & 1 \\ \hline 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- So if $\mathbf{r} = [1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1]$

$$\implies \mathbf{rH}^T = [0 \ 0 \ 0]$$

- But if $\mathbf{r} = [1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 0]$

$$\implies \mathbf{rH}^T = [0 \ 0 \ 1]$$

= Error syndrome \mathbf{s}

- Note that \mathbf{s} is the last row of \mathbf{H}^T

- Also note error took place in the last bit

- Syndrome indicates error position



Linear Block codes

- Cyclic code

➤ A code $C = \{c_1, c_2, \dots, c_{2^k}\}$ is cyclic if

$$(c_1, c_2, \dots, c_n) \in C \quad \Rightarrow \quad (c_n, c_1, \dots, c_{n-1}) \in C$$

➤ (7,4) Hamming code is cyclic

message	codeword
0001	0001101
1000	1000110
0100	0100011



Linear Block codes

- Important parameters

Hamming Distance between codewords c_i and c_j :

$$d(c_i, c_j) = \# \text{ of components at which the 2 codewords differ}$$

Hamming weight of a codeword c_i is

$$w(c_i) = \# \text{ of non-zero components in the codeword}$$

Minimum Hamming Distance of a code:

$$d_{\min} = \min d(c_i, c_j) \text{ for all } i \neq j$$

Minimum Weight of a code:

$$w_{\min} = \min w(c_i) \text{ for all } c_i \neq 0$$

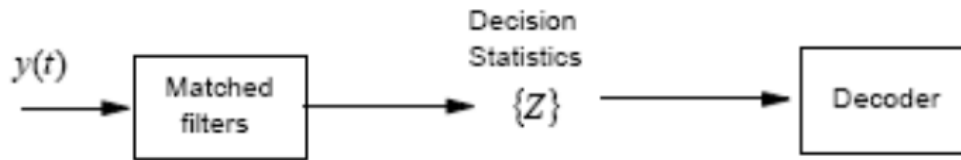
Theorem: In any linear code, $d_{\min} = w_{\min}$

Exercise: Find d_{\min} for (7,4) Hamming code



Linear Block codes

- Soft-decision and hard-decision decoding
 - **Soft-decision** decoder operates directly on the decision statistics



- **Hard-decision** decoder makes “hard” decision (0 or 1) on individual bits



- Here we only focus on hard decision decoder



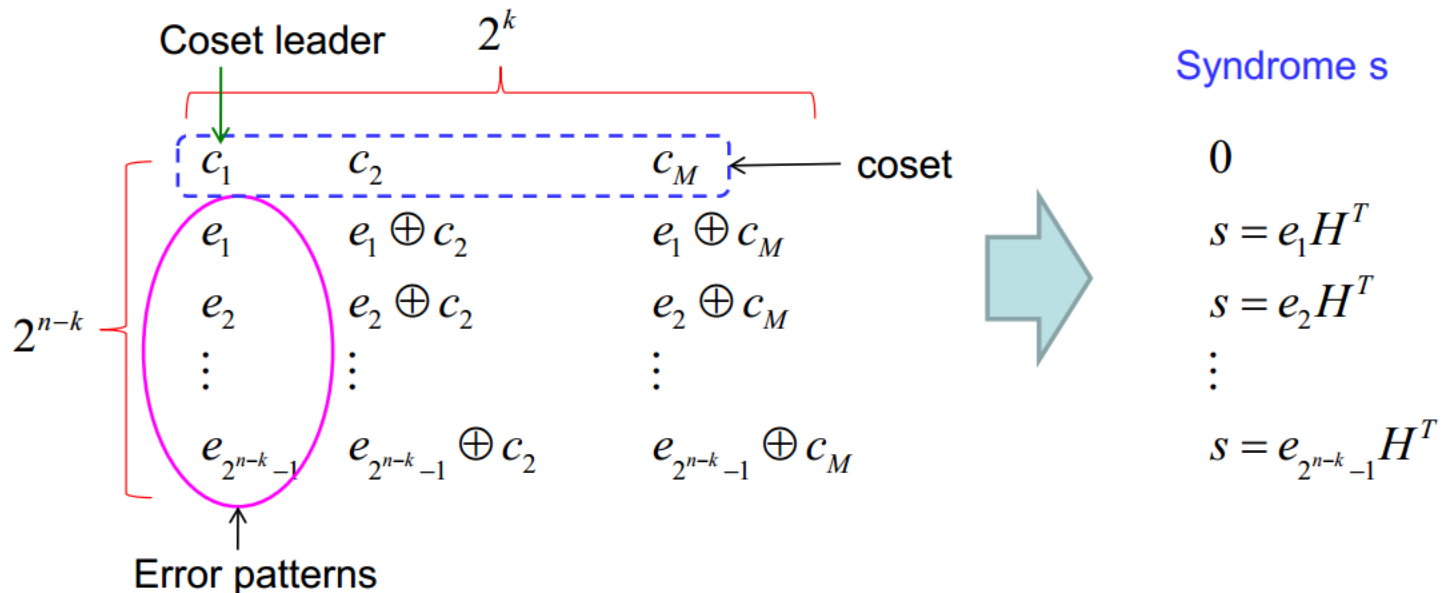
Linear Block codes

- Hard-decision decoding
 - **Minimum Hamming distance decoding**
 1. Given the received codeword \mathbf{r} , choose \mathbf{c} which is closest to \mathbf{r} in terms of Hamming distance
 2. To do so, one can do an exhaustive search (but complexity problem if k is large)
 - **Syndrome decoding**
 1. Syndrome testing: $\mathbf{r}=\mathbf{c}+\mathbf{e}$ with $\mathbf{s}=\mathbf{r}\mathbf{H}^T$
 2. This implies that the corrupted codeword \mathbf{r} and the error pattern have the same syndrome
 3. A simplified decoding procedure based on the above observation can be used



Linear Block codes

- Hard-decision decoding
 - Let the codewords be denoted as $\{c_1, c_2, \dots, c_M\}$ with c_1 being the all-zero codeword
 - A standard array is constructed as





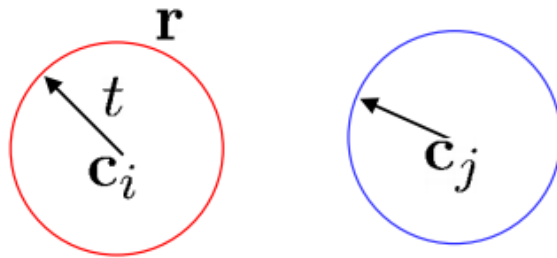
Linear Block codes

- Hard-decision decoding
 - Hard-decoding procedure
 1. Find the syndrome by \mathbf{r} using $\mathbf{s}=\mathbf{rH}^T$
 2. Find the coset corresponding to \mathbf{s} by using the standard array
 3. Find the cost leader and decode as $\mathbf{c}=\mathbf{r}+\mathbf{e}_j$
 - Try on (7,4) Hamming code

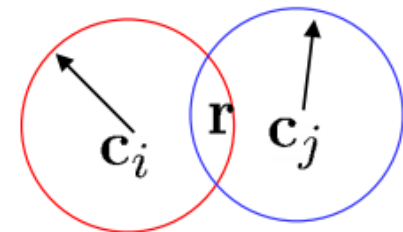


Linear Block codes

- Hard-decision decoding
 - A linear block code with a minimum distance d_{\min} can
 1. **Detect** up to $(d_{\min}-1)$ errors in each codeword
 2. **Correct** up to $t = \lfloor \frac{d_{\min} - 1}{2} \rfloor$ errors in each codeword
 3. t is known as the **error correction capability** of the codeword



$$d(\mathbf{c}_i, \mathbf{c}_j) \geq 2t + 1$$



$$d(\mathbf{c}_i, \mathbf{c}_j) < 2t$$



Linear Block codes

- Hard-decision decoding
 - Consider a linear block code (n,k) with an error correcting capability t . The decoder can correct all combination of errors up to and including t errors
 - Assume that the error probability of each individual coded bit is p and that bit errors occur independently since the channel is memoryless
 - If we send n -bit block, the probability of receiving a specific pattern of m errors and $(n-m)$ correct bits is

$$p^m(1-p)^{n-m}$$

- Total number of distinct patterns of n bits with m errors and $(n-m)$ correct bits is

$$\binom{n}{m} = \frac{n!}{m!(n-m)!}$$



Linear Block codes

- Hard-decision decoding

- Total probability of receiving a pattern with m errors is

$$P(m, n) = \binom{n}{m} \cdot p^m (1 - p)^{n-m}$$

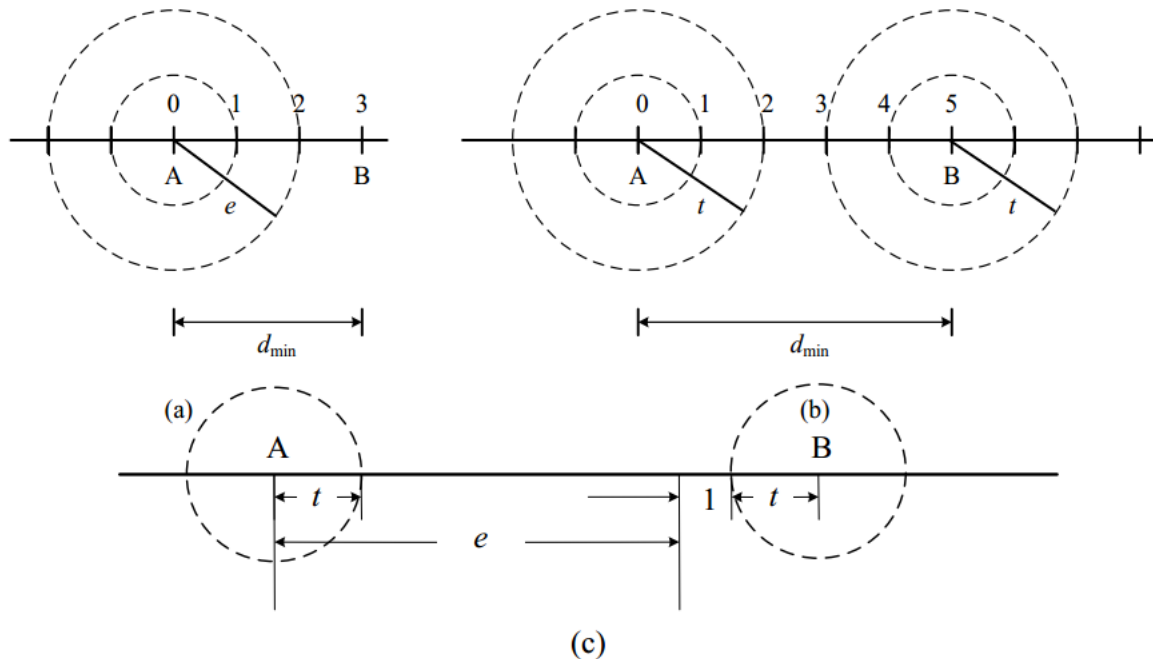
- Thus, the codeword error probability is upperbounded by

$$P_M \leq \sum_{m=t+1}^n \binom{n}{m} p^m (1 - p)^{n-m}$$

(with equality for perfect codes)

Linear Block codes

- Hard-decision decoding
 - Key parameters.



- To detect e bit errors, we have $d_{min} \geq e + 1$
- To correct t bit errors, we have $d_{min} \geq 2t + 1$



Linear Block codes

- Major classes of block codes
 - Repetition code
 - Hamming code
 - Golay code
 - BCH code
 - Reed-Solomon codes
 - Walsh codes
 - **LDPC codes**: invented by Robert Gallager in his PhD thesis in 1960, now proved to be capacity approaching and adopted in **5G** standards



Convolutional codes

- A convolutional code has memory
 - It is described by 3 integers: n , k , and L
 - Maps k inf. bits into n bits using previous $(L-1)k$ bits
 - The n bits emitted by the encoder are not only a function of the current input k bits, but also a function of the previous $(L-1)k$ bits
 - **Code rate** = k/n (information bits/coded bits)
 - **L** is the constraint length and is a measure of the **code memory**
 - n **does not** define a block or codeword length



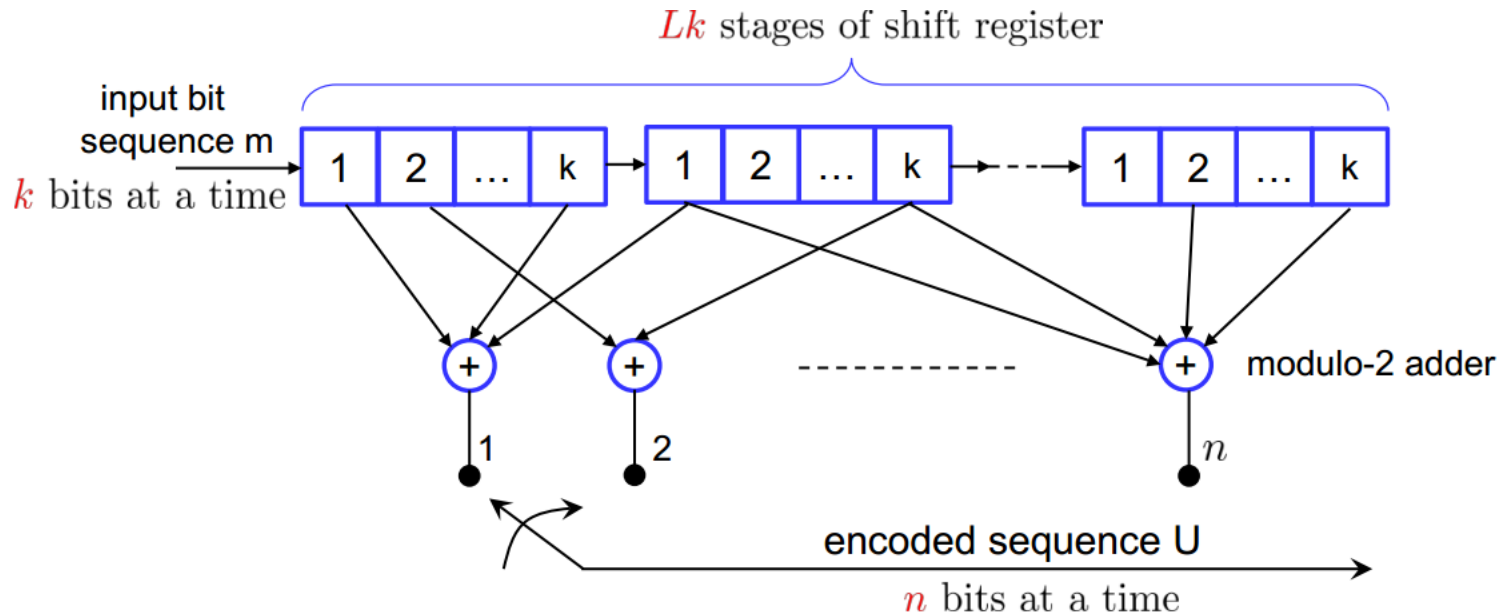
Convolutional codes

- Convolutional encoding
 - A rate k/n convolutional encoder with constraint length L consists of kL -stage shift register and n mod-2 adders
 - At each unit of time
 1. k bits are shifted into the 1st k stages of the register
 2. All bits in the register are shifted k stages to the right
 3. The output of the n adders are sequentially sampled to give the coded bits
 4. There are n coded bits for each input group of k bits or message bits. Hence $R=k/n$ information bits/coded bits is the code rate ($k < n$)



Convolutional codes

- Convolutional encoding
 - Encoder structure.



- Typically, $k=1$ for binary codes. Hence, consider rate $1/n$ codes for example.



Convolutional codes

- Convolutional encoding
 - Encoding function: characterizes the relationship between the information sequence m and the output coded sequence U .
 - Four popular methods for representation
 1. Connection pictorial and connection polynomials (usually for **encoder**)
 2. State diagram
 3. Tree diagram **Usually for decoder**
 4. Trellis diagram



Convolutional codes

- Convolutional encoding
 - Connection representation.
 - Specify n connection vectors, $g_i, i = 1, \dots, n$ for each of the n mod-2 adders
 - Each vector has kL dimension and describes the connection of the shift register to the mod-2 adders
 - A 1 in the i^{th} position of the connection vector implies shift register is connected
 - A 0 implies no connection exists



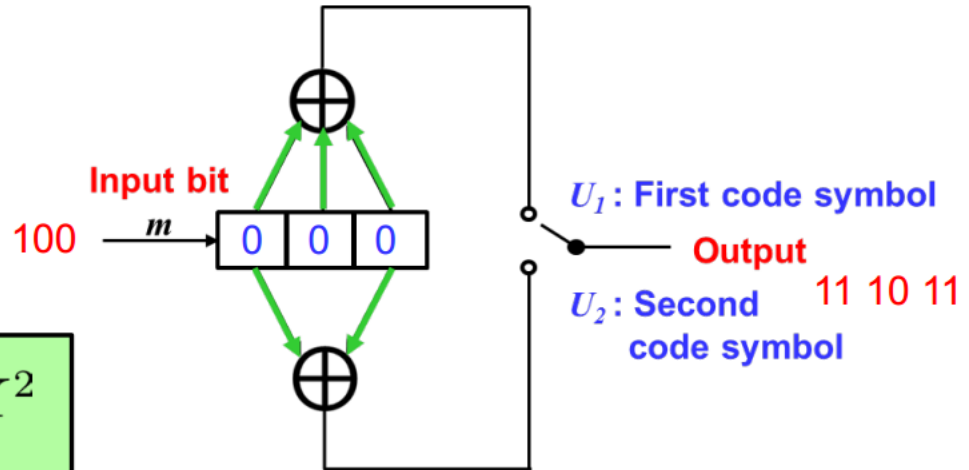
Convolutional codes

- Convolutional encoding
 - Connection representation (L=3, Rate 1/2).

$$g_1 = 111$$
$$g_2 = 101$$

Or

$$g_1(X) = 1 + X + X^2$$
$$g_2(X) = 1 + X^2$$



If Initial Register content is 0 0 0 and Input Sequence is 1 0 0. Then Output Sequence is 11 10 11



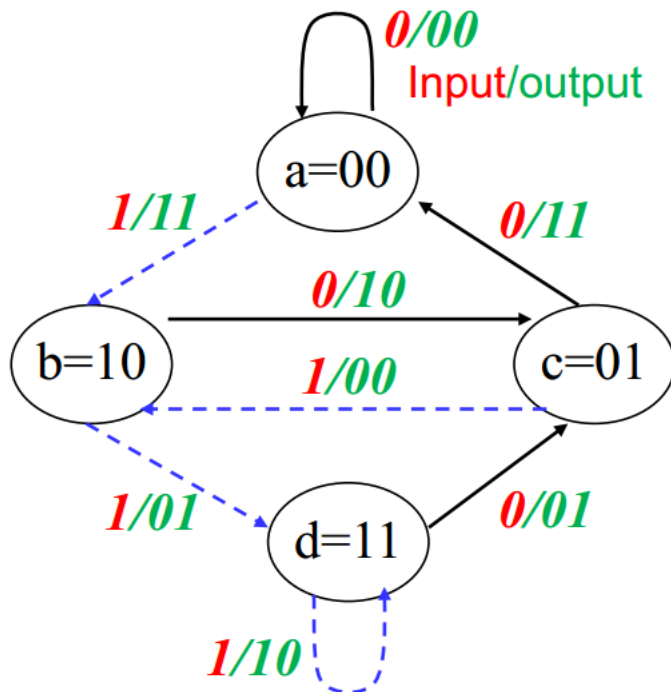
Convolutional codes

- Convolutional encoding
 - **State diagram representation.**
 - The contents of the rightmost $L-1$ stages (or the previous $L-1$ bits) are considered the current state, 2^{L-1} states
 - Knowledge of the current state and the next input is necessary and sufficient to determine the next output and next state
 - For each state, there are only 2 transitions (to the next state) corresponding to the 2 possible input bits
 - The transitions are represented by paths on which we write the output word associated with the state transition: A solid line path corresponds to an input bit 0, while dashed line for 1



Convolutional codes

- Convolutional encoding
 - State diagram representation (L=3, Rate 1/2).



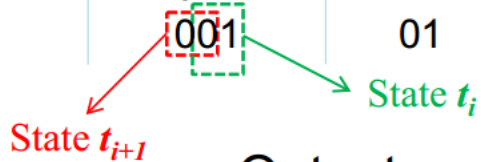
Current State	Input	Next State	Output
00	0	00	00
	1	10	11
10	0	01	10
	1	11	01
01	0	00	11
	1	10	00
11	0	01	01
	1	11	10



Convolutional codes

- Convolutional encoding
 - State diagram representation.
 - Assume that $m=11011$ is the input followed by $L-1=2$ zeros to flush the register. Also assume that the initial register contents are all zero. Find the output sequence U

Input bit m_i	Register contents	State at time t_i	State at time t_{i+1}	Branch word at time t_i	
				u_1	u_2
--	000	00	00	--	--
1	100	00	10	1	1
1	110	10	11	1	1
0	011	11	01	0	0
1	101	01	10	1	1
1	110	10	11	1	1
0	011	11	01	0	0
0	001	01	00	0	0

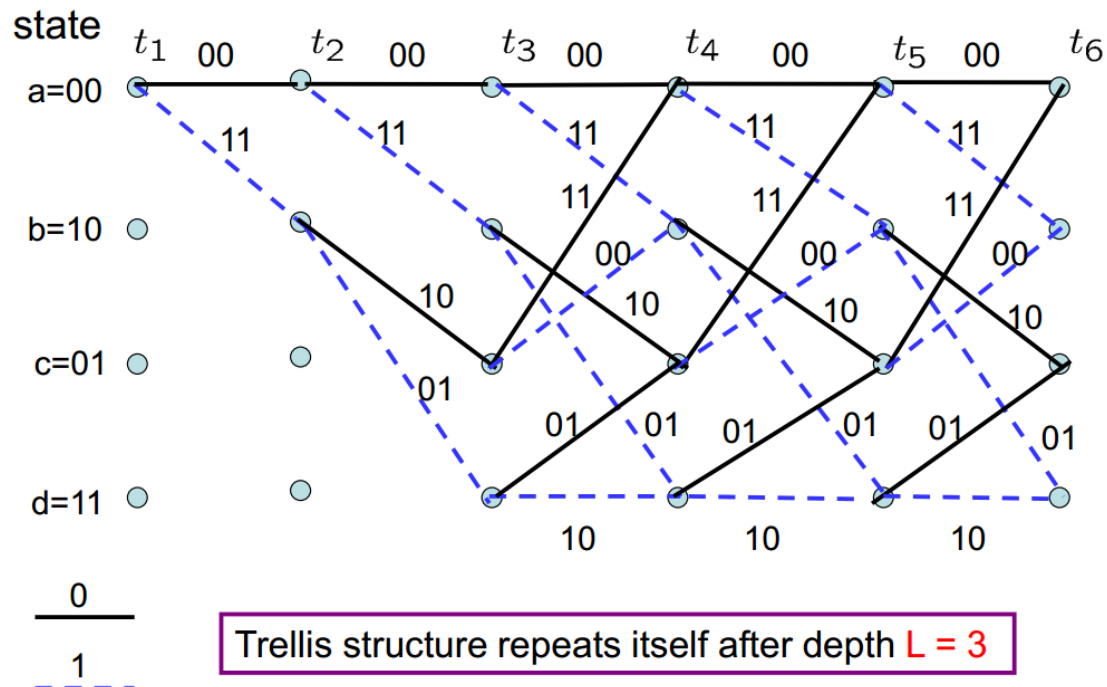


Output sequence: $U = 11\ 01\ 01\ 00\ 01\ 01\ 11$



Convolutional codes

- Convolutional encoding
 - Trellis diagram representation.
 - Trellis diagram is similar to the state diagram, except that it adds the dimension of time.
 - The code is represented by a trellis where each trellis branch describes an output word



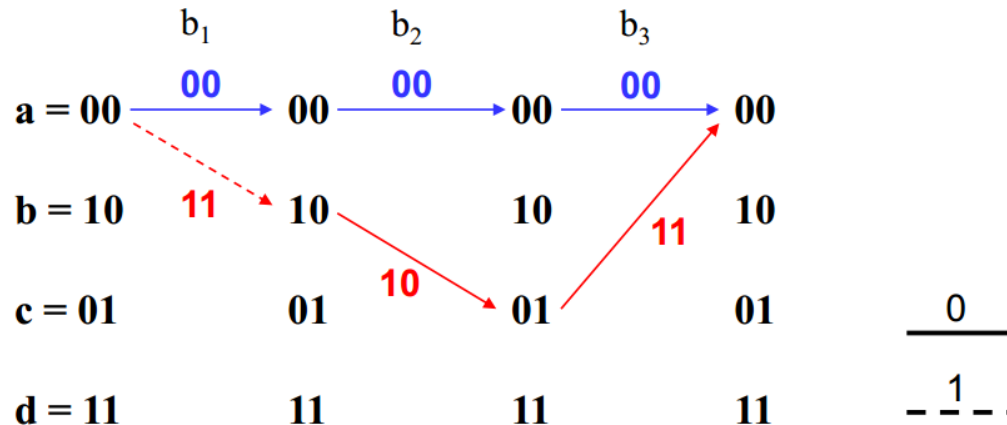


Convolutional codes

- Convolutional encoding
 - Trellis diagram representation.
 - Every input sequence (m_1, m_2, \dots) corresponds to
 1. a path in the trellis
 2. a state transition sequence (s_0, s_1, \dots) , (assume $s_0=0$ is fixed)
 3. an output sequence $((u_1, u_2), (u_3, u_4), \dots)$
 - For instance, let $s_0=00$, then

$b_1 b_2 b_3 = 000$ gives output *000000* and states *aaaa*

$b_1 b_2 b_3 = 100$ gives output *111011* and states *abca*



$\frac{0}{-}$
 $\frac{1}{-}$



Convolutional codes

- Update
 - We have discussed conv. code with constraint length L and rate $1/n$, and the different representations
 - We will talk about decoding of convolutional code with **maximum likelihood decoding**, **Viterbi algorithm**, and **transfer function**



Convolutional codes

- Maximum likelihood decoding
 - Transmit a coded sequence $U^{(m)}$ (corresponds to message sequence m) using a digital modulation scheme (e.g., BPSK or QPSK)
 - Received sequence z
 - Maximum likelihood decoder will
 1. Find the sequence $U^{(j)}$ such that
$$P(Z|U^{(j)}) = \max_{\forall U^{(m)}} P(Z|U^{(m)})$$
 2. Minimize the probability of error if m is equally likely



Convolutional codes

- Maximum likelihood decoding

- Assume a memoryless channel, i.e., noise components are independent. Then, for a rate $1/n$ code

$$P(\mathbf{Z}|\mathbf{U}^{(m)}) = \prod_{i=1}^{\infty} P(Z_i|U_i^{(m)}) = \prod_{i=1}^{\infty} \prod_{j=1}^n P(z_{ji}|u_{ji}^{(m)})$$

\downarrow
 i -th branch of \mathbf{Z}

- Then, the problem is to find a path through the trellis such that

by taking log

$$\begin{aligned} & \max_{\mathbf{U}^{(m)}} \prod_{i=1}^{\infty} \prod_{j=1}^n P(z_{ji}|u_{ji}^{(m)}) \\ & \max_{\mathbf{U}^{(m)}} \sum_{i=1}^{\infty} \sum_{j=1}^n \log P(z_{ji}|u_{ji}^{(m)}) \\ & = \max_{\mathbf{U}^{(m)}} \sum_{i=1}^{\infty} \sum_{j=1}^n LL(z_{ji}|u_{ji}^{(m)}) \end{aligned}$$

Log-likelihood path metric

i -th branch metric

Log-likelihood of $z_{ji}|u_{ji}^{(m)}$



Convolutional codes

- Maximum likelihood decoding
 - Log-likelihood.
 - For AWGN channel with soft decision

$z_{ji} = u_{ji} + n_{ji}$ and $P(z_{ji}|u_{ji})$ is Gaussian with mean u_{ji} and variance σ^2

Hence

$$\ln p(z_{ji}|u_{ji}) = -\frac{1}{2} \ln(2\pi\sigma^2) - \frac{(z_{ji} - u_{ji})^2}{2\sigma^2}$$

Note that the objective is to compare which $\sum_i \ln(p(z|u))$ for different \mathbf{u} is larger, hence, constant and scaling does not affect the results

Then, we let the log-likelihood be $LL(z_{ji}|u_{ji}) = -(z_{ji} - u_{ji})^2$

and

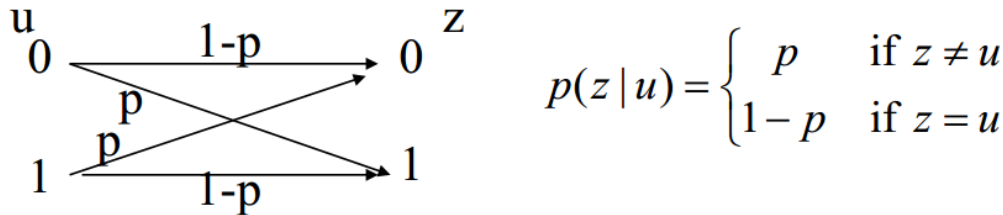
$$\log P(Z|U^{(m)}) = - \sum_{i=1}^{\infty} \sum_{j=1}^n \left(z_{ji} - u_{ji}^{(m)} \right)^2$$

Thus, soft decision ML decoder is to choose the path whose corresponding sequence is at the minimum Euclidean distance from the received sequence



Convolutional codes

- Maximum likelihood decoding
 - Log-likelihood.
 - For binary symmetric channel (hard decision)



$$LL(z_{ji} | u_{ji}) = \ln p(z_{ji} | u_{ji}) = \begin{cases} \ln p & \text{if } z_{ji} \neq u_{ji} \\ \ln(1-p) & \text{if } z_{ji} = u_{ji} \end{cases}$$

$$= \begin{cases} \ln p / (1-p) & \text{if } z_{ji} \neq u_{ji} \\ 0 & \text{if } z_{ji} = u_{ji} \end{cases}$$

$$= \begin{cases} -1 & \text{if } z_{ji} \neq u_{ji} \quad (\text{since } p < 0.5) \\ 0 & \text{if } z_{ji} = u_{ji} \end{cases}$$

▪ Thus

$$\log P(Z|U^{(m)}) = -d_m$$

Hamming distance between Z and U^(m), i.e. they differ in d_m positions

Hard-Decision ML Decoder = Minimum Hamming Distance Decoder



Convolutional codes

- Maximum likelihood decoding
 - Decoding procedure:
 1. Compute, for each branch i , the branch metric using output bits $\{u_{1,i}, u_{2,i}, \dots, u_{n,i}\}$ associated with that branch and the received symbols $\{z_{1,i}, z_{2,i}, \dots, z_{n,i}\}$
 2. Compute, for each valid path through the trellis (a valid codeword sequence $U(m)$), the sum of the branch metrics along that path
 3. The path with the maximum path metric is the decoded path
 - To compare all possible valid paths, we need to do exhaustive search or brute-force, not practical as the # of paths grows exponentially as the path length increases
 - The optimal algorithm for solving this problem is the **Viterbi decoding algorithm** or **Viterbi decoder**



Convolutional codes

- Viterbi decoding



**Andrew Viterbi
(1935-)**

- BS & MS in MIT
- PhD in University of Southern California
- Invention of Viterbi algorithm in 1967
- Co-founder of Qualcomm Inc. in 1983





Convolutional codes

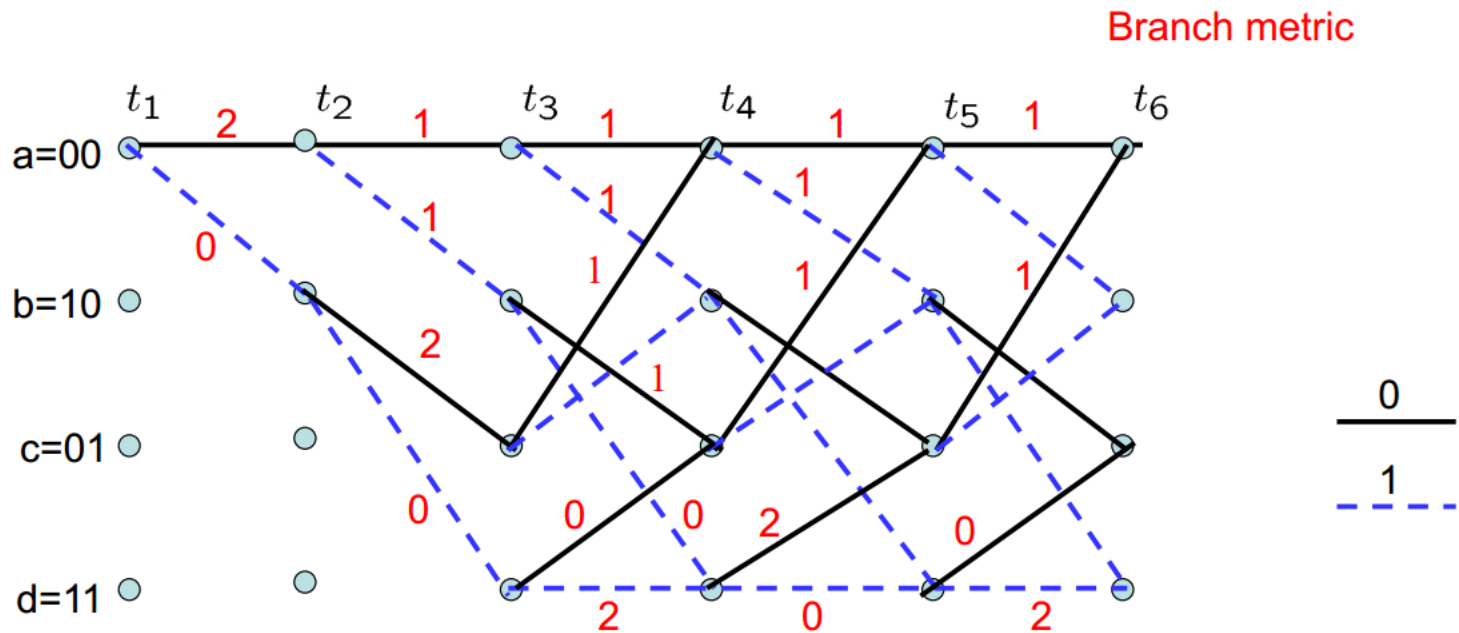
- Viterbi decoding

➤ Consider $R=1/2$, $L=3$ for example.

Input data sequence \mathbf{m} : 1 1 0 1 1 ...

Coded sequence \mathbf{U} : 11 01 01 00 01 ...

Received sequence \mathbf{Z} : 11 01 01 10 01 ...





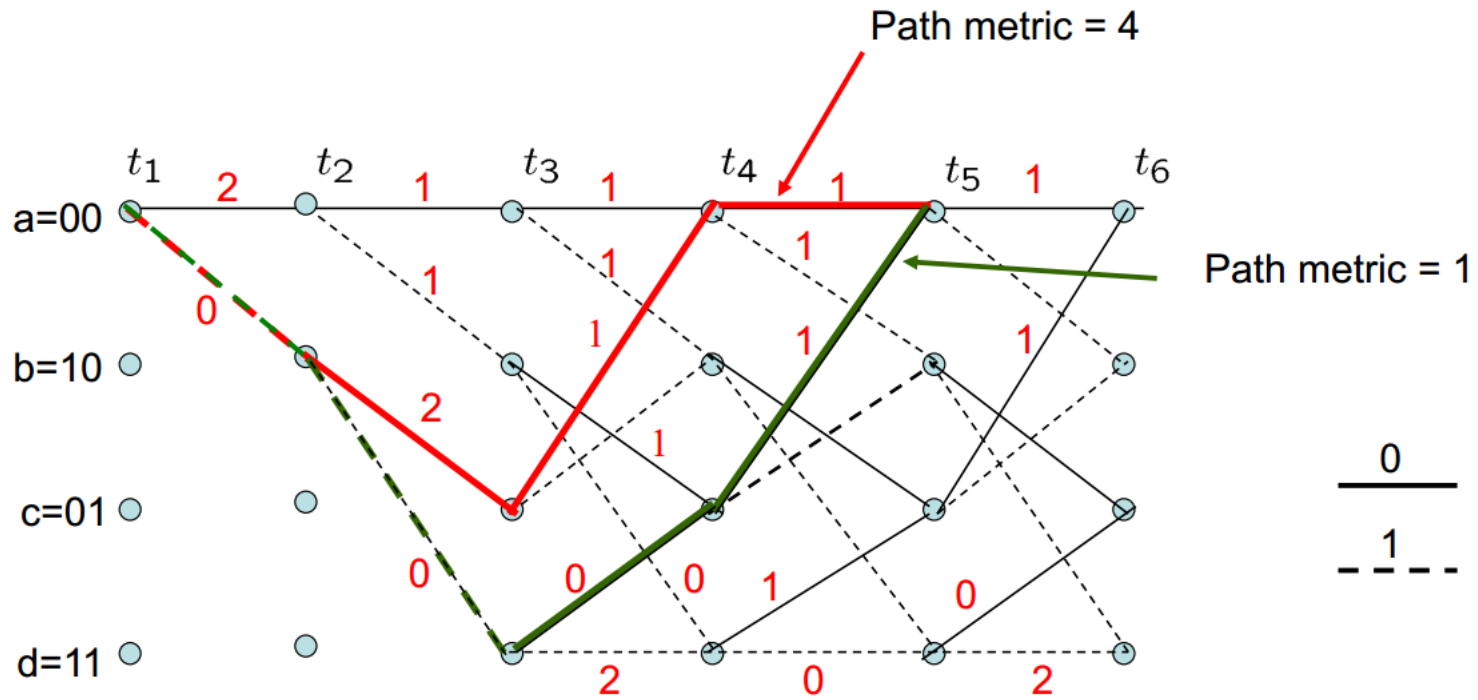
Convolutional codes

- Viterbi decoding
 - Basic idea: If any 2 paths in the trellis merge to a single state, one of them can always be eliminated in the search
 - Let cumulative path metric of a given path at t_i = sum of the branch metrics along that path up to time t_i
 - Consider t_5
 1. The upper path metric is 4, the lower path metric is 1
 2. The upper path metric cannot be path of the optima path since the lower path has a lower metric
 3. This is because future output branches depend on the current state and not the previous state



Convolutional codes

- Viterbi decoding





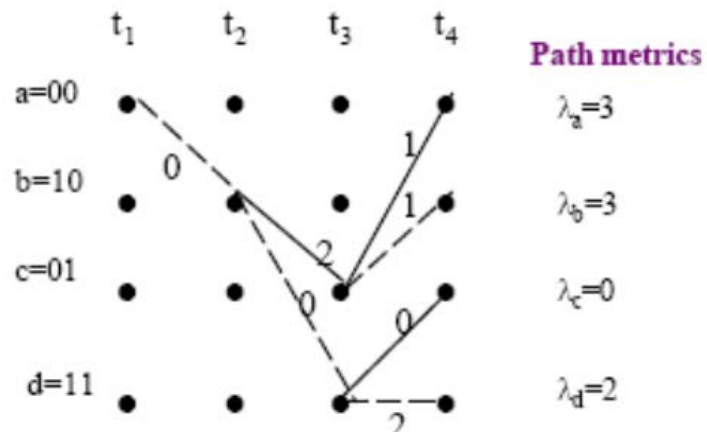
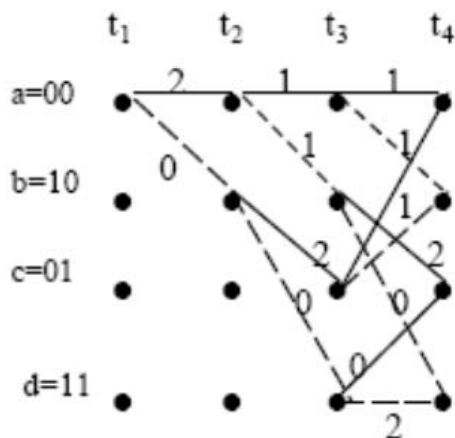
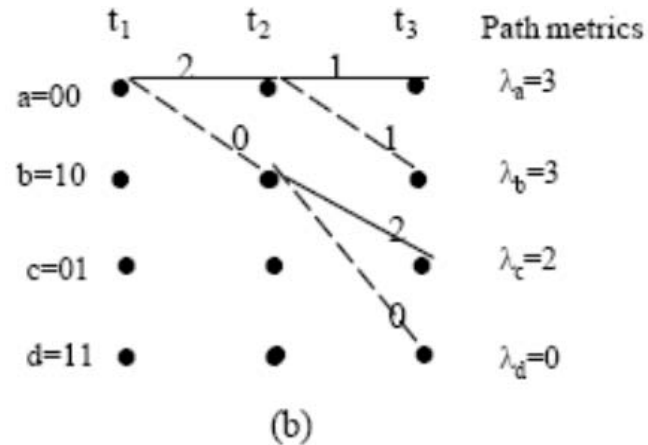
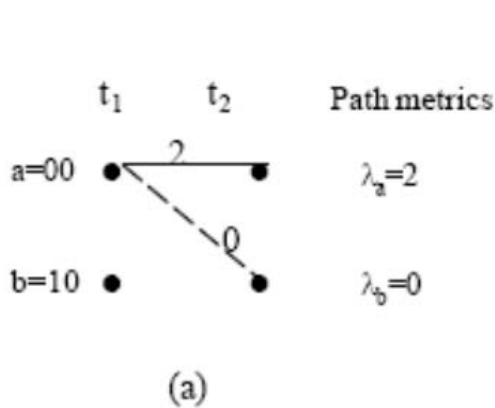
Convolutional codes

- Viterbi decoding
 - At time t_i , there are 2^L-1 states in the trellis
 - Each state can be entered by means of 2 states
 - Viterbi decoding consists of computing the metric of the 2 paths entering each state and eliminating one of them
 - This is done for each of the 2^L-1 nodes at time t_i
 - The decoder then moves to time t_{i+1} and repeat the process



Convolutional codes

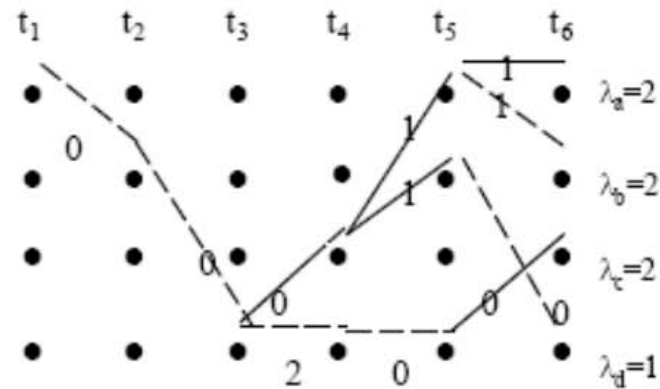
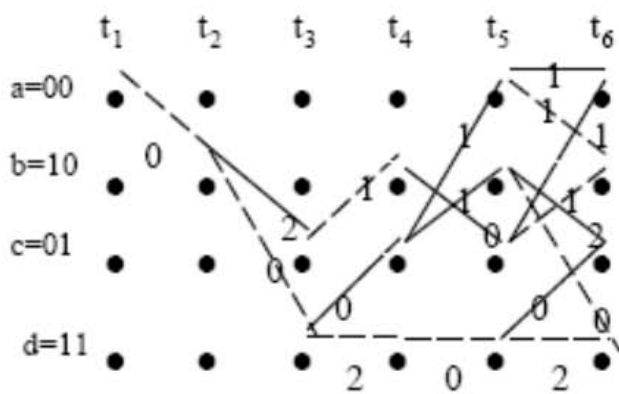
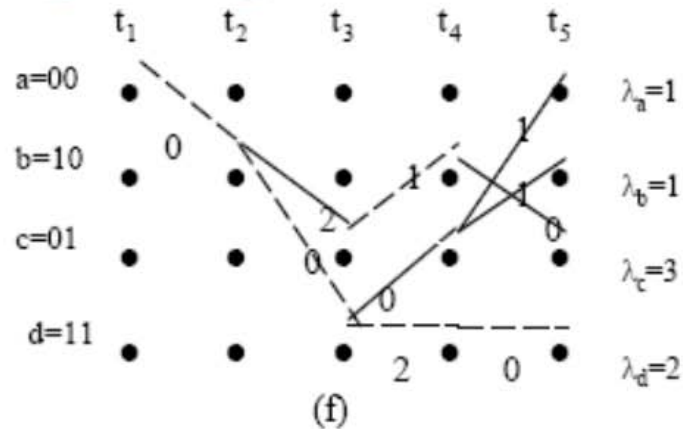
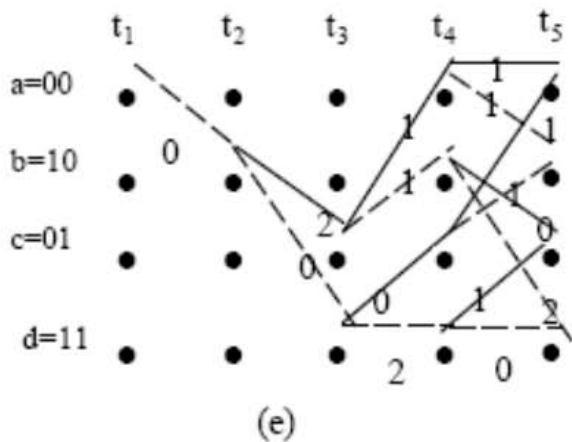
- Viterbi decoding
 - Example.





Convolutional codes

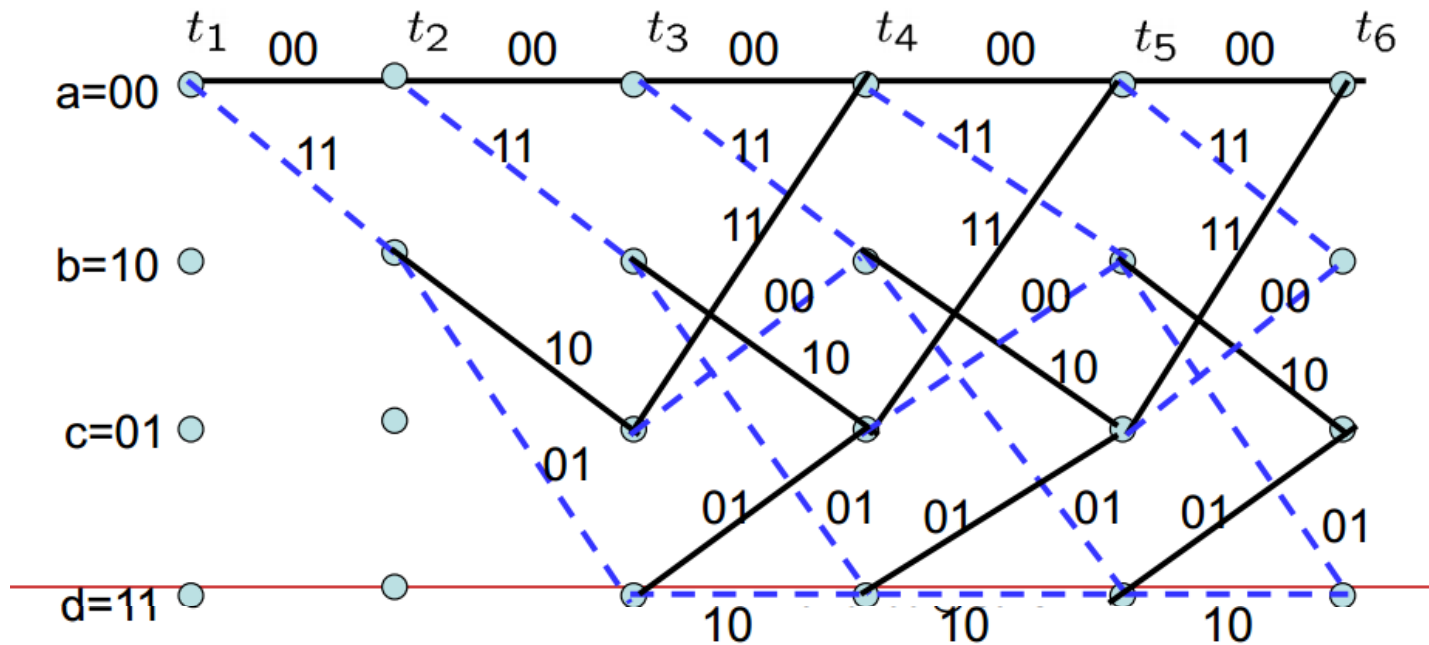
- Viterbi decoding
 - Example.





Convolutional codes

- Viterbi decoding
 - d_{free} =Minimum free distance=Minimum distance of any pair of arbitrarily long paths that diverge and remerge
 - A code can correct any t channel errors where (this is an approximation) $t \leq \lfloor \frac{d_{\text{free}} - 1}{2} \rfloor$





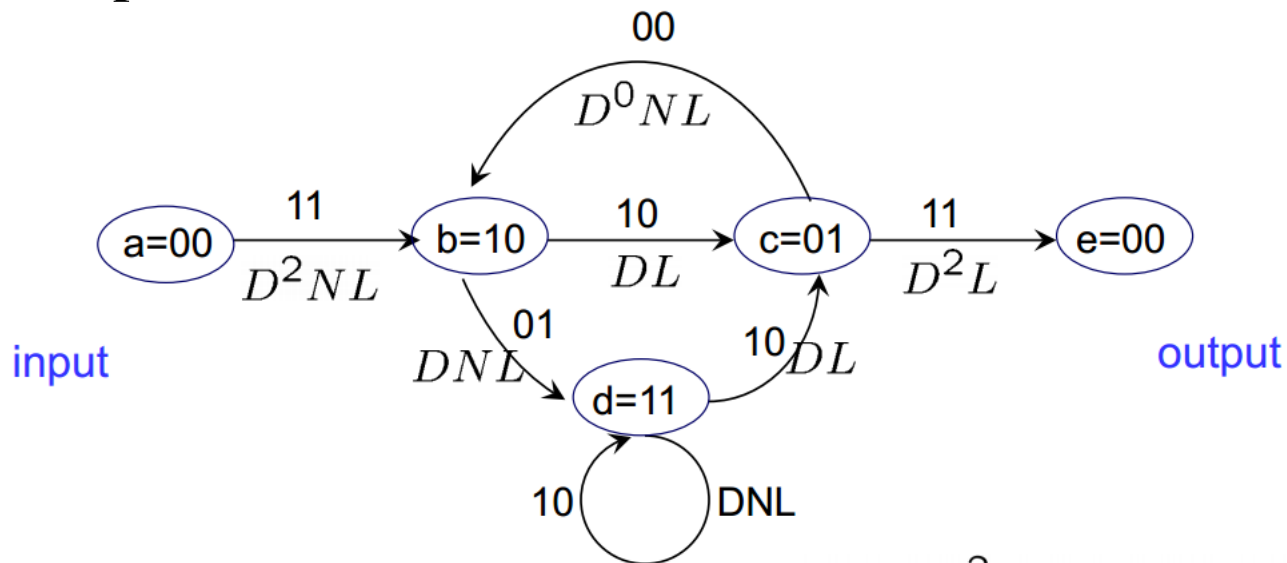
Convolutional codes

- Transfer function
 - The distance properties and the error rate performance of a convolutional code can be obtained from its transfer function
 - Since a convolutional code is linear, the set of Hamming distances of the code sequences generated up to some stages in the trellis, from the all-zero code sequence, is the same as the set of distances of the code sequences with respect to any other code sequence
 - Thus, we assume that the all-zero path is the input to the encoder



Convolutional codes

- Transfer function
 - State diagram labeled according to distance from all-zero path



D^m denote m non-zero output bits
 N if the input bit is non-zero
 L denote a branch in the path

$$\begin{cases}
 X_b = D^2NLX_a + LNX_c \\
 X_c = DLX_b + DLX_d \\
 X_d = DNLX_b + DNLX_d \\
 X_e = D^2LX_c
 \end{cases}$$



Convolutional codes

- Transfer function

- The transfer function $T(D,N,L)$, also called the weight enumerating function of the code is

$$T(D, N, L) = \frac{X_e}{X_a}$$

- By solving the state equations we get

$$\begin{aligned} T(D, N, L) &= \frac{D^5NL^3}{1 - DNL(1 + L)} \\ &= D^5NL^3 + D^6N^2L^4(1 + L) + D^7N^3L^5(1 + L)^2 \\ &\quad + \dots + D^{l+5}N^{l+1}L^{l+3}(1 + L)^l + \dots \end{aligned}$$

- The transfer functions indicates that

1. There is one path at distance 5 and length 3, which differs 1 bit from the correct all-zeros path
2. There are 2 paths at distance 6, one of which is of length 4, the other length 5, and both differ in 2 input bits from all-zeros path
3. $d_{\text{free}} = 5$



Convolutional codes

- Good convolutional codes
 - Good convolutional codes can only be found in general by computer search
 - They are listed in tables and classified by their constraint length, code rate, and their generator polynomials or vectors (typically using octal notation).
 - The error-correction capability of a convolutional code increases as n increases or as the code rate decreases.
 - Thus, the channel bandwidth and decoder complexity increases.



Convolutional codes

- Good convolutional codes
 - Rate 1/2.

Constraint Length	Generator Polynomials	d_{free}
3	(5,7,7)	8
4	(13,15,17)	10
5	(25,33,37)	12
6	(47,53,75)	13
7	(133,145,175)	15
8	(225,331,367)	16
9	(557,663,711)	18
10	(1117,1365,1633)	20



Convolutional codes

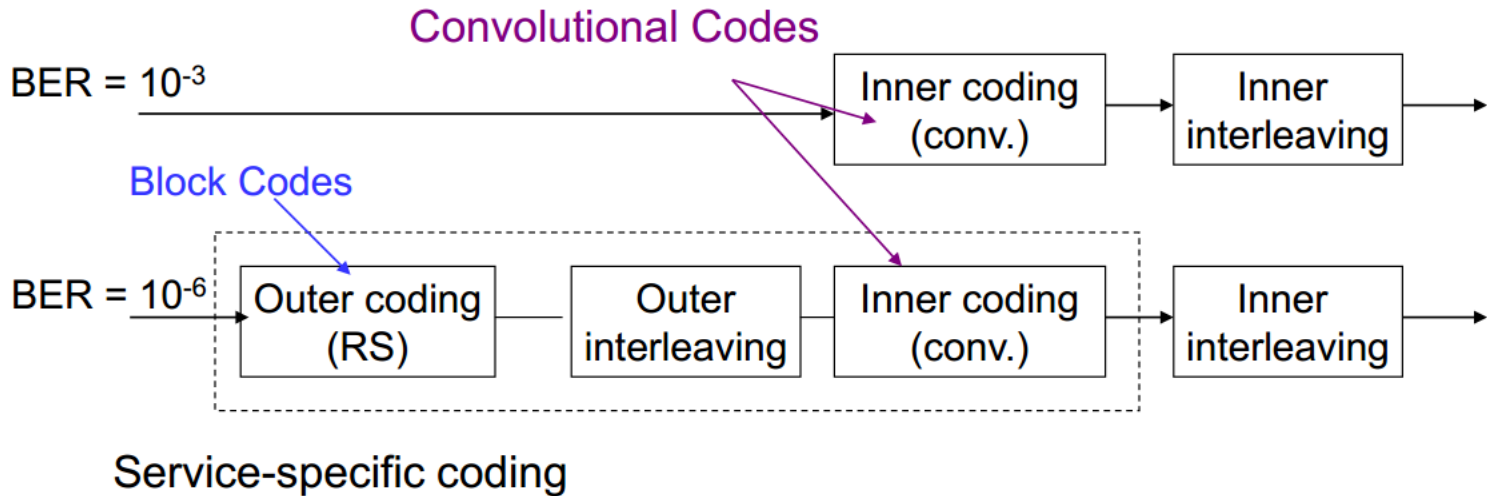
- Good convolutional codes
 - Rate 1/3.

Constraint Length	Generator Polynomials	d_{free}
3	(5,7)	5
4	(15,17)	6
5	(23,35)	7
6	(53,75)	8
7	(133,171)	10
8	(247,371)	10
9	(561,753)	12
10	(1167,1545)	12



Convolutional codes

- Channel coding for Wideband CDMA



Convolutional code is rate $1/3$ and rate $1/2$,
all with constraint length 9



Convolutional codes

- Channel coding for Wireless LAN (IEEE 802.11a)

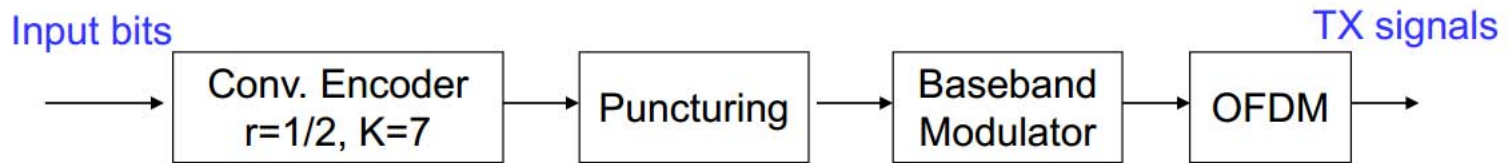


Table 11-3. Encoding details for different OFDM data rates

Speed (Mbps)	Modulation and coding rate (R)	Coded bits per carrier ^[a]	Coded bits per symbol	Data bits per symbol ^[b]
6	BPSK, R=1/2	1	48	24
9	BPSK, R=3/4	1	48	36
12	QPSK, R=1/2	2	96	48
18	QPSK, R=3/4	2	96	72
24	16-QAM, R=1/2	4	192	96
36	16-QAM, R=3/4	4	192	144
48	64-QAM, R=2/3	6	288	192
54	64-QAM, R=3/4	6	288	216



Convolutional codes

- Other advanced channel coding
 - Low density parity check codes: Robert Gallager 1960
 - Turbo codes: Berrou et al. 1993
 - Trellis-coded modulation: Ungerboeck 1982
 - Space-time coding: Vehid Tarokh et al. 1998
 - Polar codes: Erdal Arkan 2009

Check the latest coding techniques in 5G standards