

EE 577 - Wireless and Personal Communications

Chapter 7: Mitigation Techniques

Fading Mitigation Techniques

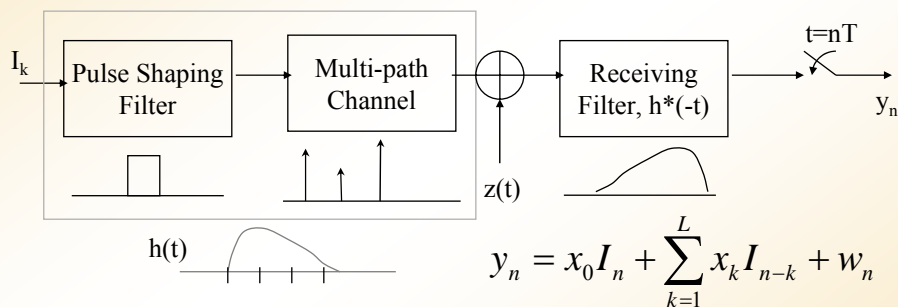
- Equalization:** mitigates frequency-selective fading channels
- Diversity:** mitigates flat fading
- Channel Coding:** mitigates errors due to fading and noise, provide time diversity
- Adaptive Modulation:** mitigates time-selective fading

Causes of ISI

- ❑ Channel Distortion
- ❑ **Multipath Fading:** Frequency selective channels act as an FIR filter causing channel-induced ISI
- ❑ Pulse shapes not designed for zero ISI (GMSK used in GSM)

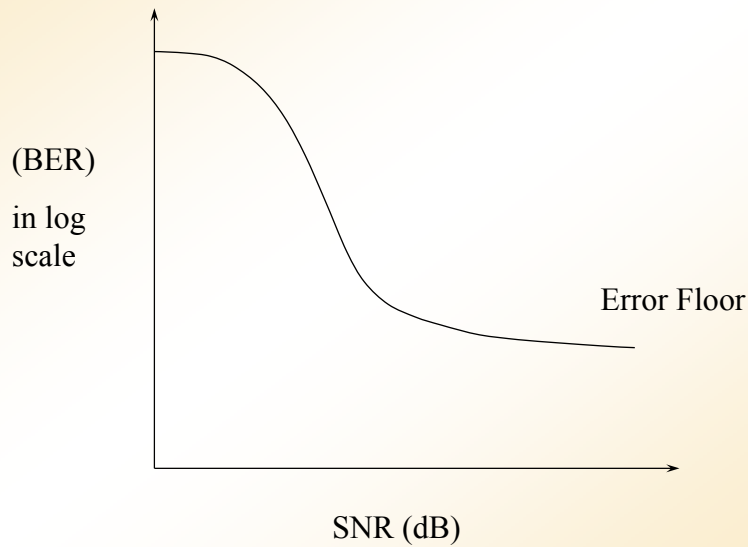
Problems in FS Fading Channels

- ❑ Multi-path Channel model causes ISI at receiver



- ❑ As signal power $S \uparrow$, ISI power \uparrow
- ❑ Hence, $S/(I+N)$ decreases slowly with increasing S/N , independent of the signal power

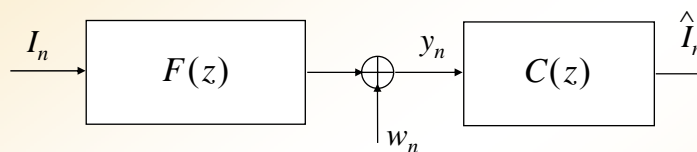
Error Floor in FS Fading Channels



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Equalization



- ❑ An equalizer is a filter that equalizes the effect of the nonideal (frequency-selective) channel
- ❑ $C(f)H(f) = 1$ (Equalizer inverses the effect of the channel)
- ❑ For time-varying channels the equalizer has to be adaptive.

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Equalizer Modes

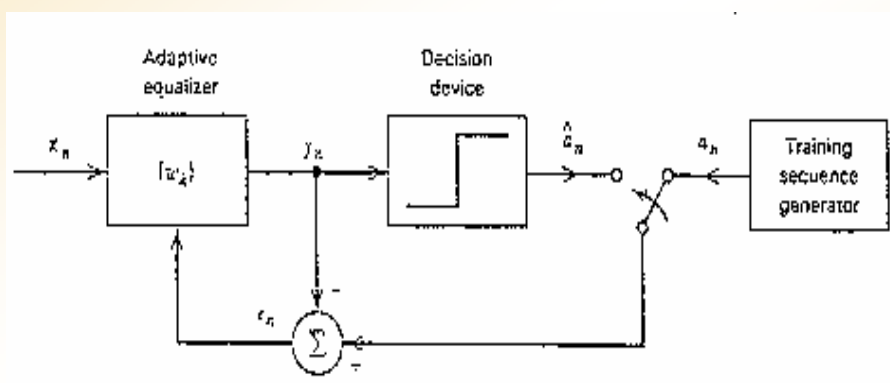
❑ Training:

- ❑ The equalizer must periodically learn the channel by transmitting a training sequence
- ❑ The training sequence is used at the receiver to choose the equalizer parameters

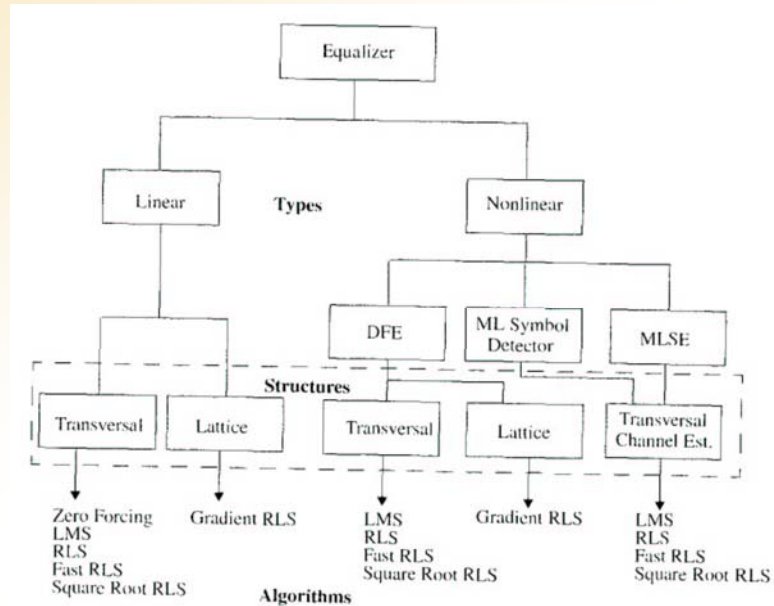
❑ Tracking:

The equalizer parameters are adjusted based on the difference between the equalizer output and the output of the decision device

Training & Decision Modes



Classification of Equalizers

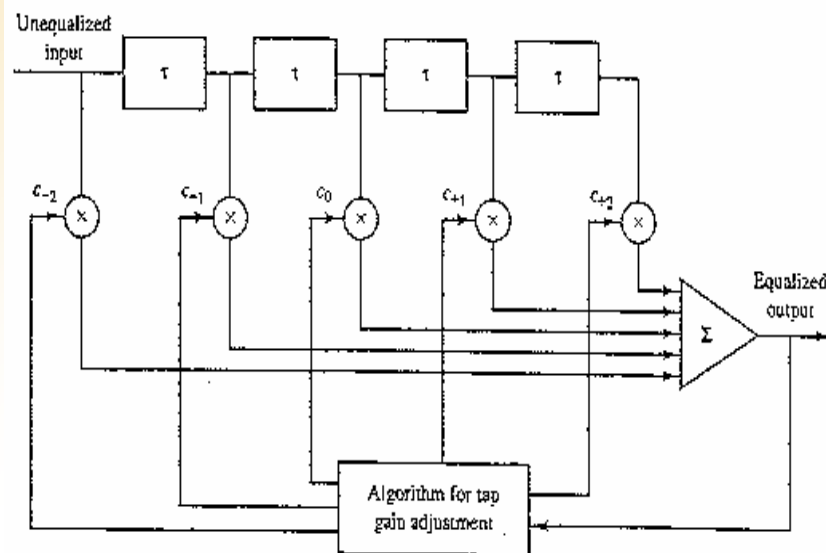


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Classification of Equalizers

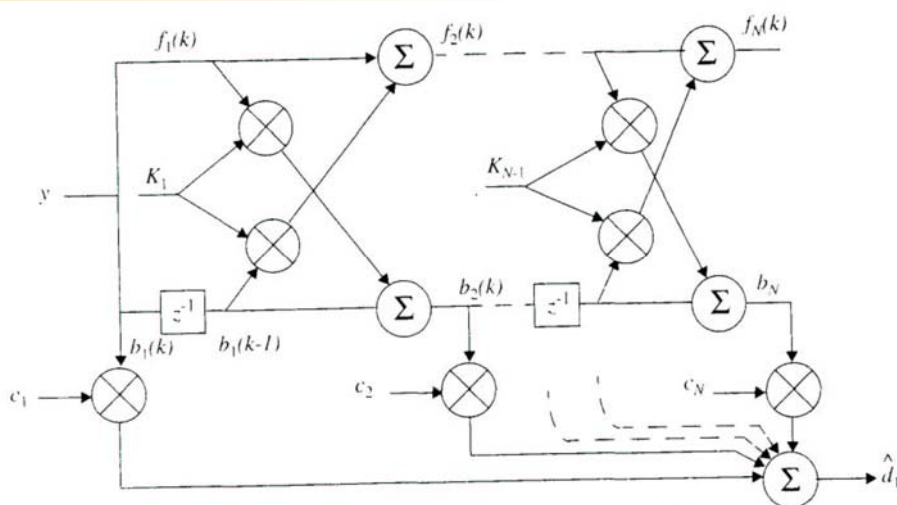
- ❑ **Linearity:**
 - ❑ Linear
 - ❑ Nonlinear (DFE, MLSE)
- ❑ **Structure:**
 - ❑ Transversal
 - ❑ Lattice
- ❑ **Algorithm:**
 - ❑ Zero Forcing
 - ❑ MMSE

Linear Transversal Equalizer



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Lattice Implementation



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Linear Equalizer

□ Design $C(z)$ or c_n such that the output is as **close** to I_n as possible

□ **Zero-Forcing:**

□ Forces the samples of combined channel/equalizer impulse response to be zero at all but one of the NT_s spaced samples

□ It is an impulse in time domain:

$$F(z)C(z) = 1$$

□ Simple, but may enhance the noise

Linear Equalizer

Minimum Mean Square Error (MMSE):

$$\begin{aligned} MSE &= E[e_k^2] = E[(x_k - \hat{x}_k)^2] \\ &= E[x_k^2] - 2p^T c + c^T R c \end{aligned}$$

$$y_k = [y_k \quad y_{k-1} \quad \cdots \quad y_{k-N}]^T$$

$$c = [c_0 \quad c_1 \quad \cdots \quad c_N]^T$$

$$\mathbf{p} = E[x_k \mathbf{y}_k^T] \quad \mathbf{R} = E[\mathbf{y}_k \mathbf{y}_k^T]$$

- To minimize MSE,

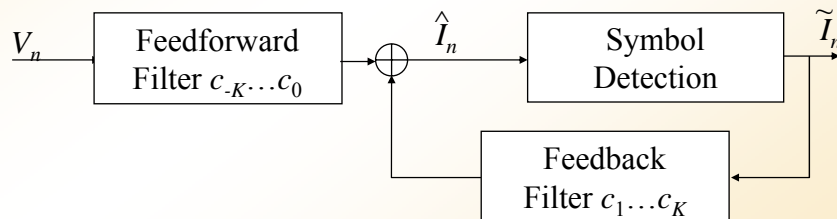
$$\frac{\partial E[e_k^2]}{\partial \mathbf{c}} = -2\mathbf{p} + 2\mathbf{R}\mathbf{c} = 0$$

$$\Rightarrow \mathbf{c}_{opt} = \mathbf{R}^{-1} \mathbf{p}$$

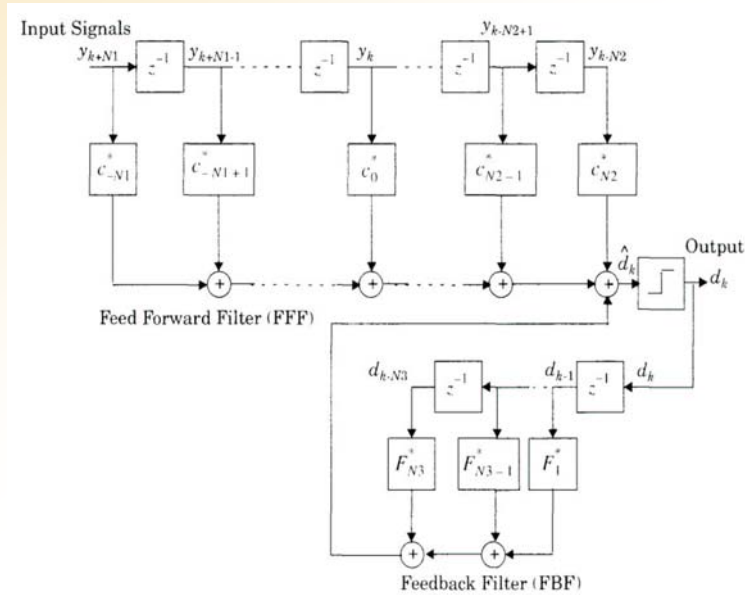
- To estimate \mathbf{R} and \mathbf{p} , the transmitter can transmit a training sequence that is known by the receiver.
- Equalizer requires periodic retraining in order to maintain effective ISI cancellation.

Decision Feedback Equalizer (DFE)

- DFE attempts to subtract from the current symbol the ISI created by previously detected symbols
- It performs better than linear equalizer always
- Suitable to channels with deep frequency nulls



DFE



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DFE

- ❑ Consists of a feed-forward filter followed by a feedback filter with the bit decisions as input
- ❑ DFE does not suffer from noise enhancements
- ❑ Coefficients can be updated using LMS, RLS or MMSE algorithms

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Performance Measures

□ Rate of convergence:

iterations required for the equalizer to converge to the correct solution

□ Misadjustment:

Measure of the amount of deviation from the correct solution

□ Computational complexity:

operations required to make one iteration in the equalizer

Adaptive Algorithms

□ Least Mean Square (LMS):

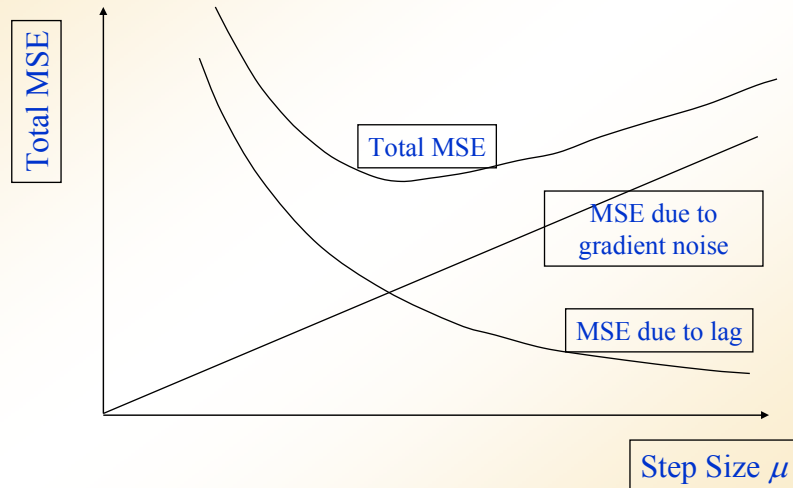
- To minimize: $MSE = E[e_k^2] = E[(x_k - \hat{x}_k)^2]$

- The algorithm is: $\mathbf{C}^{(k+1)} = \mathbf{C}^{(k)} - \mu \nabla^{(k)}$

where $\nabla^{(k)} = \frac{\partial E[e_k^2]}{\partial \mathbf{c}} = -2\mathbf{p} + 2\mathbf{R}\mathbf{c} = -2E[e_k \mathbf{y}_k]$

- So, $\mathbf{C}^{(k+1)} = \mathbf{C}^{(k)} + \mu e_k \mathbf{y}_k$

MSE in LMS Algorithm



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Training Algorithms Tradeoffs

- ❑ **LMS:** $2N+1$ multiply operations, low complexity, slow convergence, poor tracking
- ❑ **MMSE:** $N^2 - N^3$ multiply operations, very high complexity, very fast convergence, good tracking
- ❑ **RLS:** $2.5N^2 + 4.5N$ multiply operations, high complexity, fast convergence, good tracking

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Maximum Likelihood Sequence Equalizer (MLSE)

- ❑ ISI introduces some form of memory (relation between adjacent samples over the span of ISI)
- ❑ Instead of detecting the received stream symbol-by-symbol like in previously discussed equalizers
- ❑ MLSE observes a sequence of received symbols and searches for the most likely transmitted sequence

MLSE

- ❑ Comparing the received sequence to all possible transmitted sequences is a very computational complex task
- ❑ An efficient algorithm of finding the most likely sequence without the need for comprehensive search is known as the **Viterbi Algorithm**
- ❑ MLSE is used in GSM

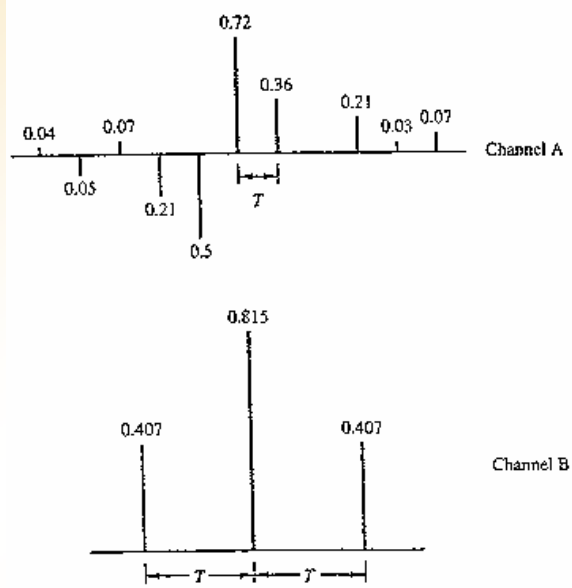
Fractionally-Spaced Equalizers (FSE)

- ❑ In previous equalizers, taps are separated by symbol duration T_s
- ❑ However, the pulse often extends to more than a symbol duration (such as in RC pulses)
- ❑ In this case FSE performs better
- ❑ In FSE, the taps are separated in time by the reciprocal of Nyquist rate ($<T_s$)
- ❑ FSE has better performance

Equalizer Design

- ❑ Complexity has to be justified by SNR gain and battery savings
- ❑ Coherence time should be greater than equalizer convergence time
- ❑ Maximum number of resolvable multipath components in the channel dictates the number of taps in the equalizer
- ❑ An equalizer can equalize a channel with a maximum delay spread less than or equal the maximum delay in the equalizer

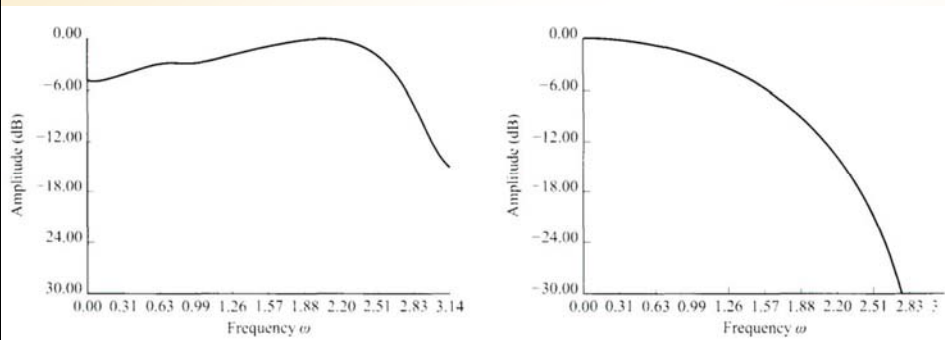
Case Study



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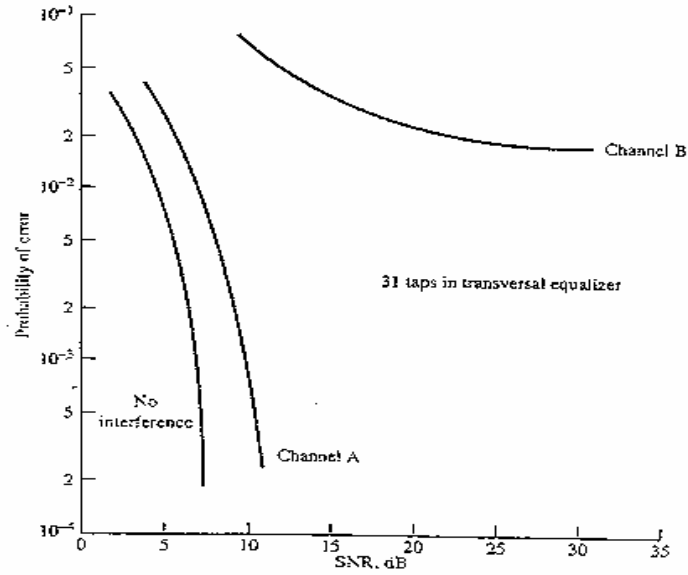
Study Case



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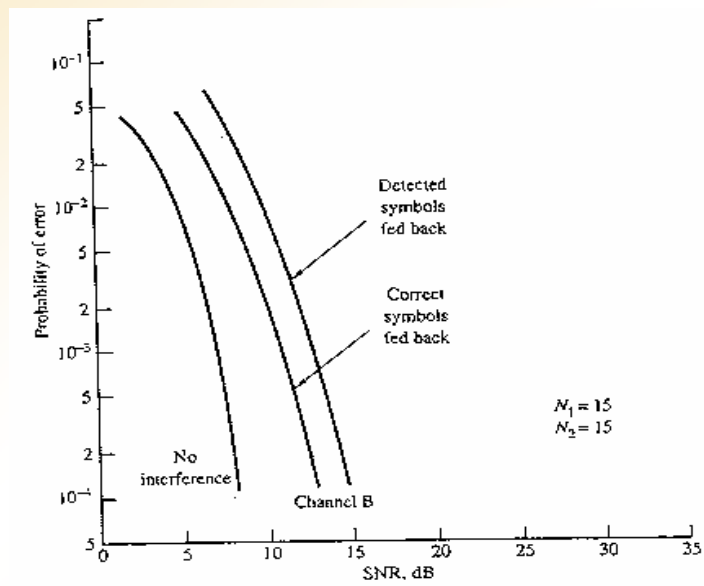
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Performance of Linear Equalizers



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Performance of DFE

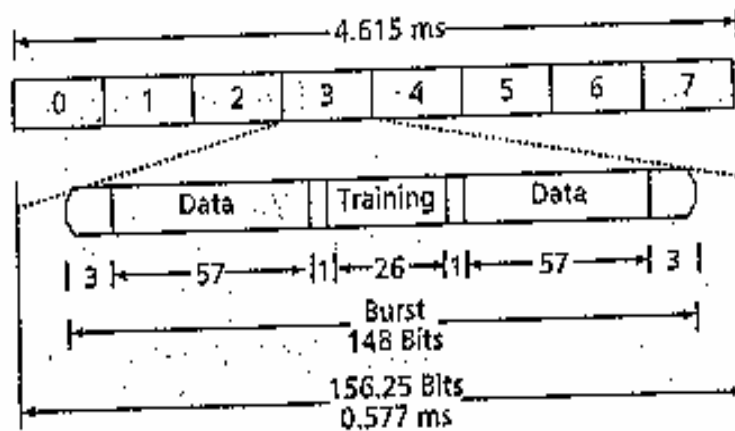


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The GSM System

- ❑ Slot duration = 0.577 ms
- ❑ Bit duration: 3.69 μ s
- ❑ Carrier freq. = 900 MHz
- ❑ $W = 200$ kHz
- ❑ $T_m = 16$ μ s
- ❑ **Assumption:**
Speed 100 km/hr (55.56 m/s)

Case Study: The GSM System



The GSM System

- ❑ Channel parameters:

$T_c \approx 3 \text{ ms}$ \Rightarrow slowly fading

$B_c \approx 62.5 \text{ kHz}$ \Rightarrow frequency selective

- ❑ Equalization is required

- ❑ Observation interval = 4-6 bits

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Lecture 16: Diversity

Diversity

- ❑ **Basic idea:** send the same information over independently fading paths, then combine the paths
- ❑ If diversity branches are uncorrelated, the probability of deeply faded received signal is reduced

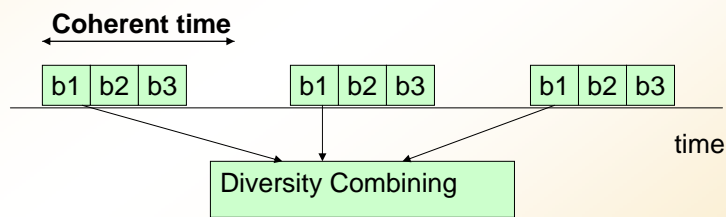
- ❑ **Macro Diversity:** provides a method to mitigate the effects of shadowing
- ❑ **Micro Diversity:** provides a method to mitigate the effects of multi-path fading

Diversity Approaches

- ❑ **Space Diversity:**
 - ❑ Using antennas spaced enough (at Tx or Rx)
- ❑ **Polarization Diversity:**
 - ❑ Using antennas with different polarizations
- ❑ **Frequency Diversity:**
 - ❑ Using frequency channels separated in frequency more than the channel coherence BW
- ❑ **Time Diversity:**
 - ❑ Using time slots separated in time more than the channel coherence time
- ❑ **Multi-path Diversity:**
 - ❑ Utilized efficiently in CDMA using RAKE receiver

Time Diversity

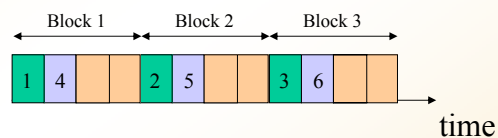
Interleave the repeated bits over a duration longer than the coherence time T_c



Block Interleaving

Example:

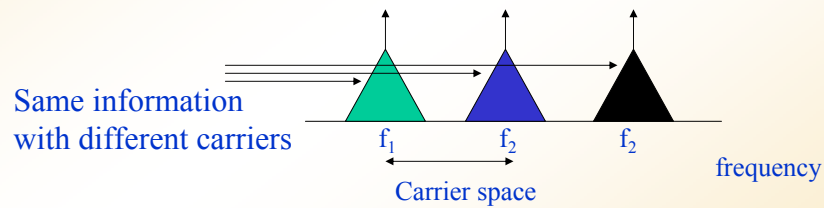
- ❑ If bit stream is 1 2 3 4 ...12
- ❑ After block interleaving,



- ❑ Less affected by burst errors

Frequency Diversity

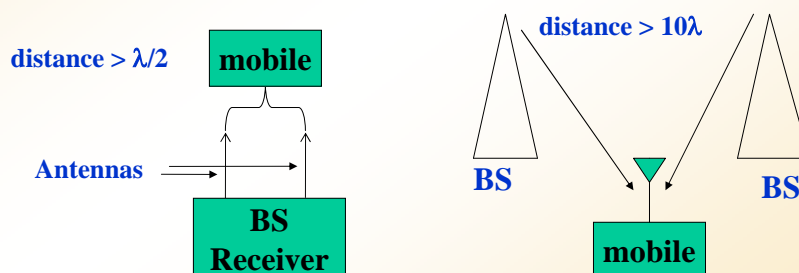
- Send the signal over multiple carriers separated in frequency by more than the **Coherence Bandwidth**



- FH-SS is a special case of frequency diversity

Space (Antenna) Diversity

- Use more than one antenna to receive the signal
- The distance between two antennas should:
 - exceed $\lambda/2$ at MS due to large amount of scatterers
 - around 10λ at BS due to less number of scatterers
- Very suitable for base station implementation



Polarization Diversity

- Use the same antenna to receive the signal
- Orthogonal polarizations (vertical and horizontal) are used to provide two diversity paths
- Reflection coefficients for vertical and horizontal polarized waves are different
- This causes orthogonally polarized waves to undergo uncorrelated fading
- Very suitable for fixed wireless links such as microwave links

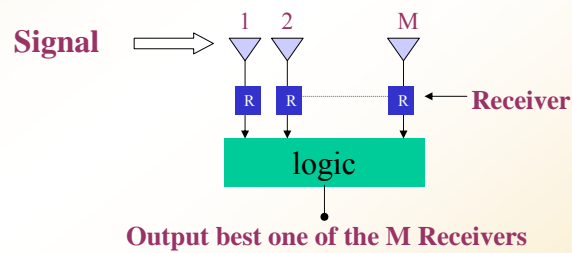
Combining Techniques

- Selection Combining (SC)
- Equal Gain Combining (EGC)
- Maximal Ratio Combining (MRC)
- Generalized SC (GSC)

MRC, EGC and GSD require a coherent phase reference to co-phase the different branch signals

Selection Combining

Key idea: Monitor **ALL** M branches at a time and select the branch with highest SNR to receive the signal



Selection Combining

- Let the average SNR in branch i be:

$$\bar{\gamma}_i = \bar{\gamma}$$

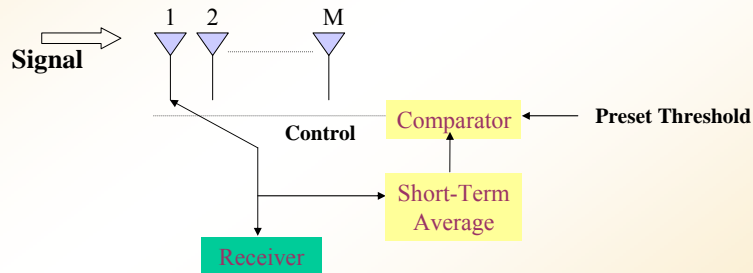
- The average SNR for Selection Combining is:

$$\bar{\gamma}_S = \bar{\gamma} \sum_{i=1}^M \frac{1}{i}$$

- The incremental gain becomes extremely small for large value of M

Scanning Diversity (Switched) Combining

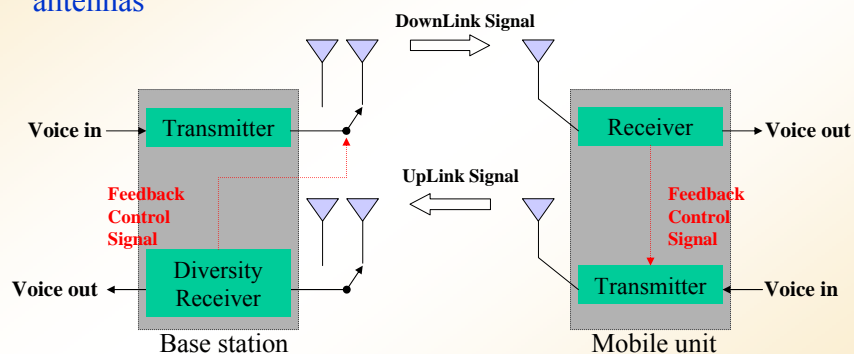
- ❑ Monitor **ONE** branch at a time.
- ❑ If signal quality of monitored branch falls below a threshold, the receiver scans other branches for better signal quality



- ❑ If threshold is large, the scanning process will be activated often
- ❑ If threshold is small, no improvement in diversity combining

Feedback Diversity Combining

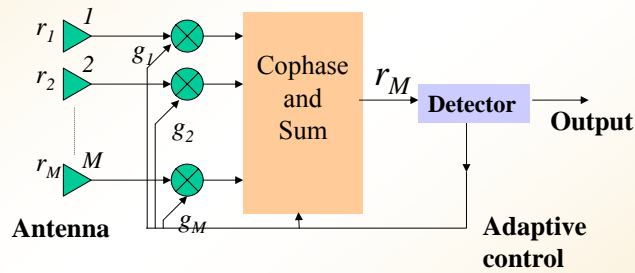
- ❑ Antenna switching is done at the BS
- ❑ Transmitter antennas are switched instead of the receiver antennas



- ❑ **Advantage:** Simplify the circuit complexity of mobile unit
- ❑ **Disadvantage:** It is not an optimal diversity technique

Maximal Ratio Combining (MRC)

- The signal from all branches are weighted and then summed together



- where s is the transmitted signal
- n_i is the i -th noise process and g_i 's are the weights

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MRC

- The signals from each of the M branches are co-phased

- The resultant signal is represented as: $r_T = \sum_{i=1}^M g_i r_i$

- Let the average SNR in branch i be: $\bar{\gamma}_i = \frac{r_i^2}{2N_i}$

- The SNR can be written as: $\gamma = \frac{1}{2} \frac{\left(\sum_{i=1}^M g_i r_i \right)^2}{\sum_{i=1}^M g_i^2 N_i}$

- Maximization gives: $g_i = \frac{r_i}{N_i}$

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MRC

❑ Thus, the SNR becomes: $\gamma_{MR} = \sum_{i=1}^M \gamma_i$

❑ The mean SNR is given by: $\bar{\gamma}_{MR} = M\bar{\gamma}$

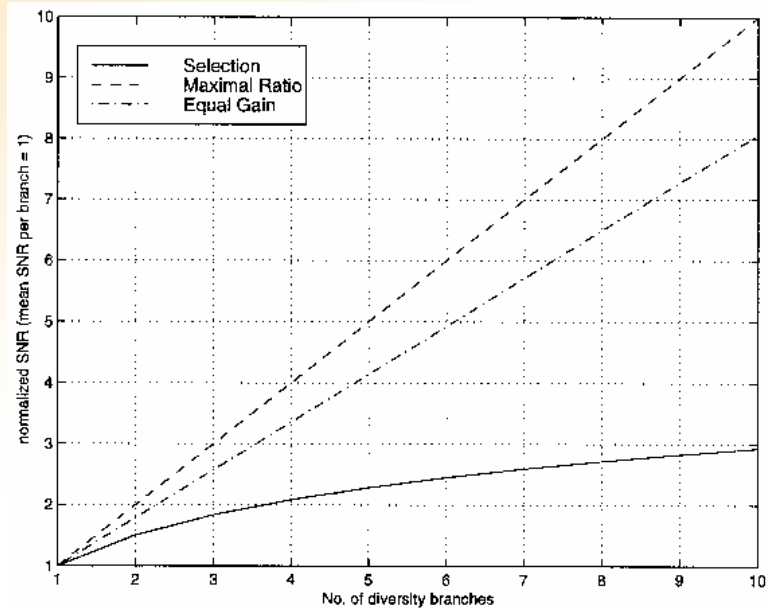
- ❑ **Advantage:** Produce an output with an acceptable SNR even when none of the individual branches are themselves acceptable
- ❑ **Disadvantage:** Channel estimation is required for each diversity gain

Equal Gain Combining (EGC)

- ❑ Similar to MRC with the weights a_i 's are all equal to 1
- ❑ No need to estimate the channel gains for each diversity branch
- ❑ The average SNR is given by:

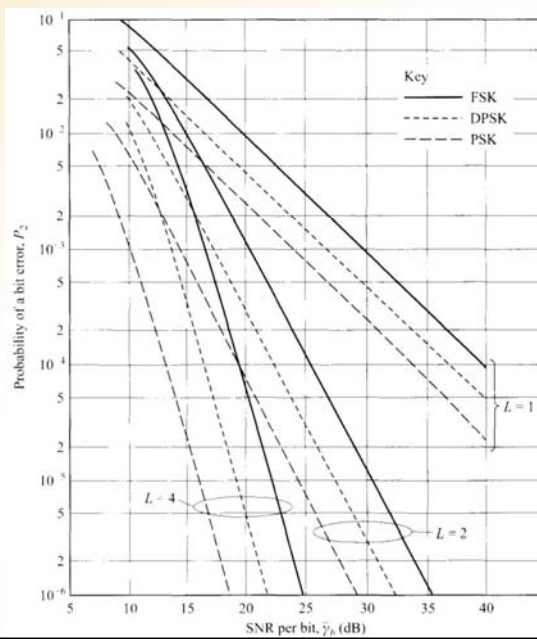
$$\bar{\gamma}_E = \bar{\gamma} \left(1 + \frac{(M-1)\pi}{4} \right)$$

Diversity SNR Gains



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Diversity Performance



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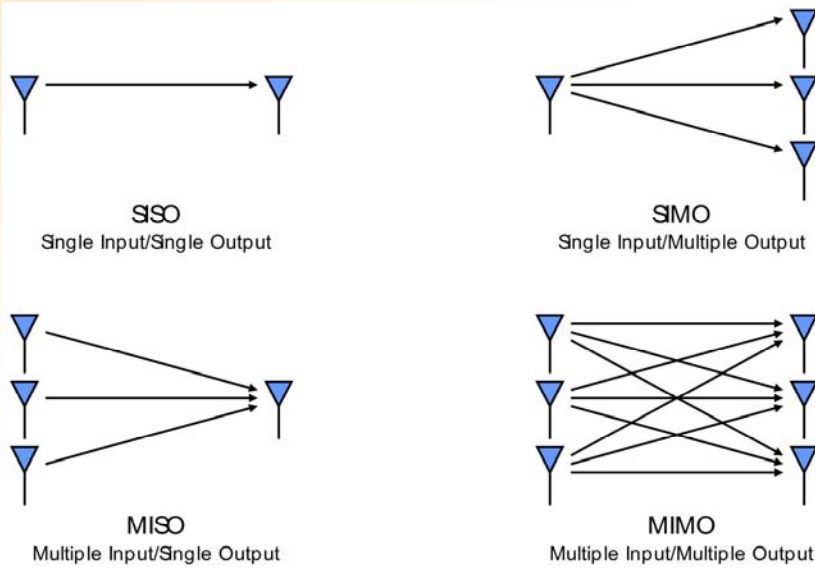
Generalized Selection Combining (GSC)

- ❑ Select the L diversity branches with the largest receive signal level (including noise and interference) among the M branches
- ❑ Combine the selected branches using MRC
- ❑ Provides a tradeoff between SC and MRC:
 - ❑ Performs better than SC
 - ❑ Less complexity than MRC
- ❑ Avoid noisy branches with small SNR values

RAKE Receiver

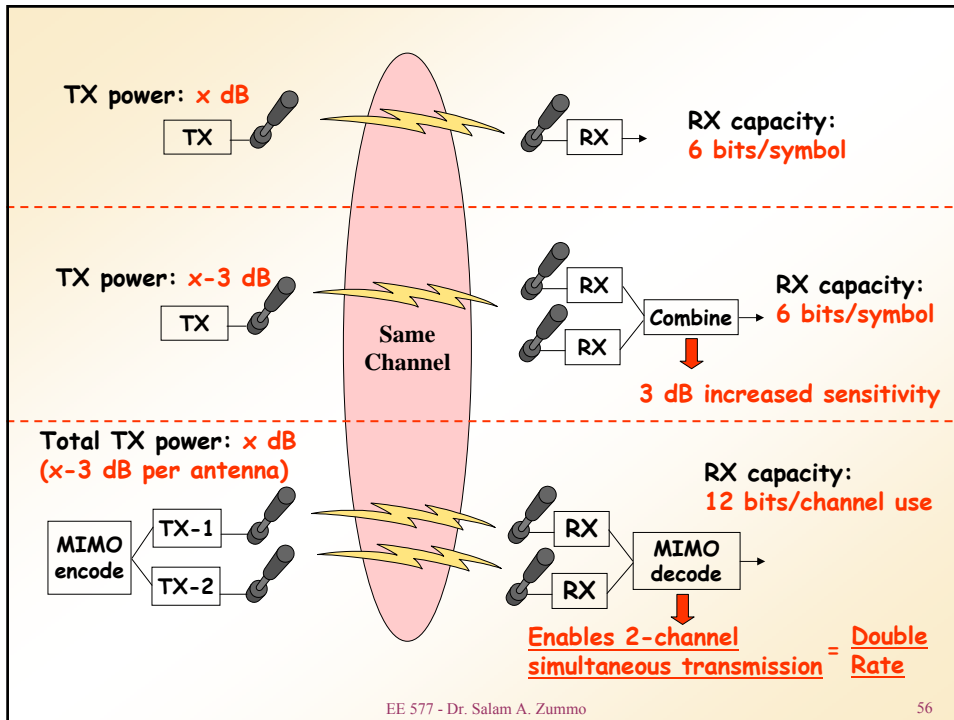
- ❑ Used with DSSS systems
- ❑ Provides a means to combine resolvable multipath components as diversity branches
- ❑ The RAKE receiver works as follows:
 - ❑ Correlate the received signal with the PN sequence
 - ❑ Correlate with a delayed version of the PN sequence to capture the first delayed multipath finger
 - ❑ Repeat delay-and-correlate process until all multipath figures are captured
 - ❑ Combine the outputs of the correlators using MRC
- ❑ **Advantage:** Takes advantage of multipath
- ❑ **Disadvantage:** Needs several correlators

Multi-Input Multi-Output (MIMO)



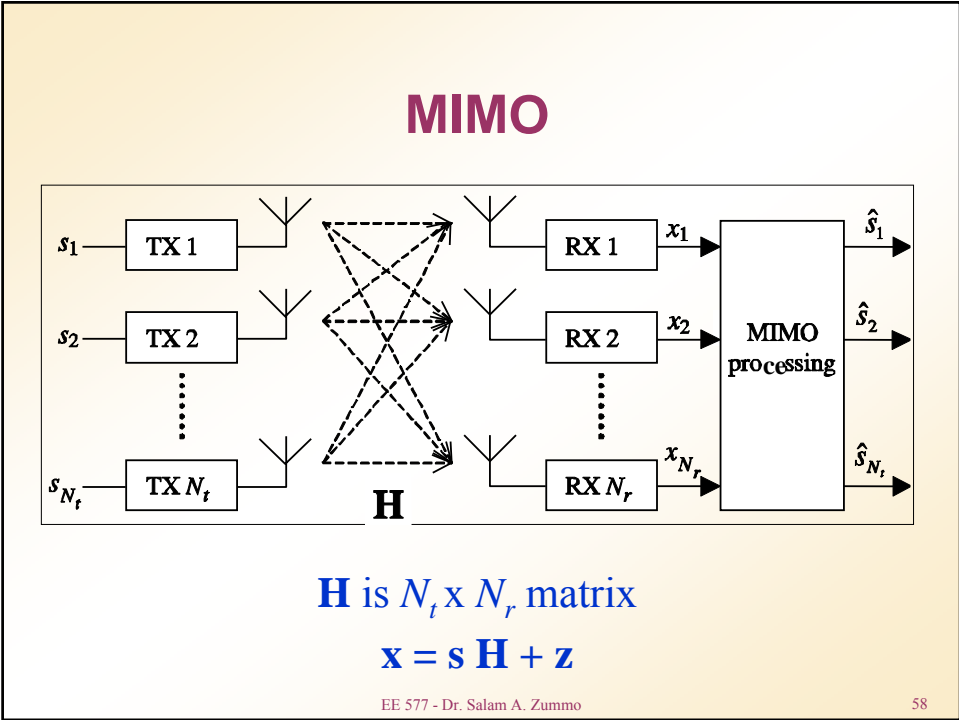
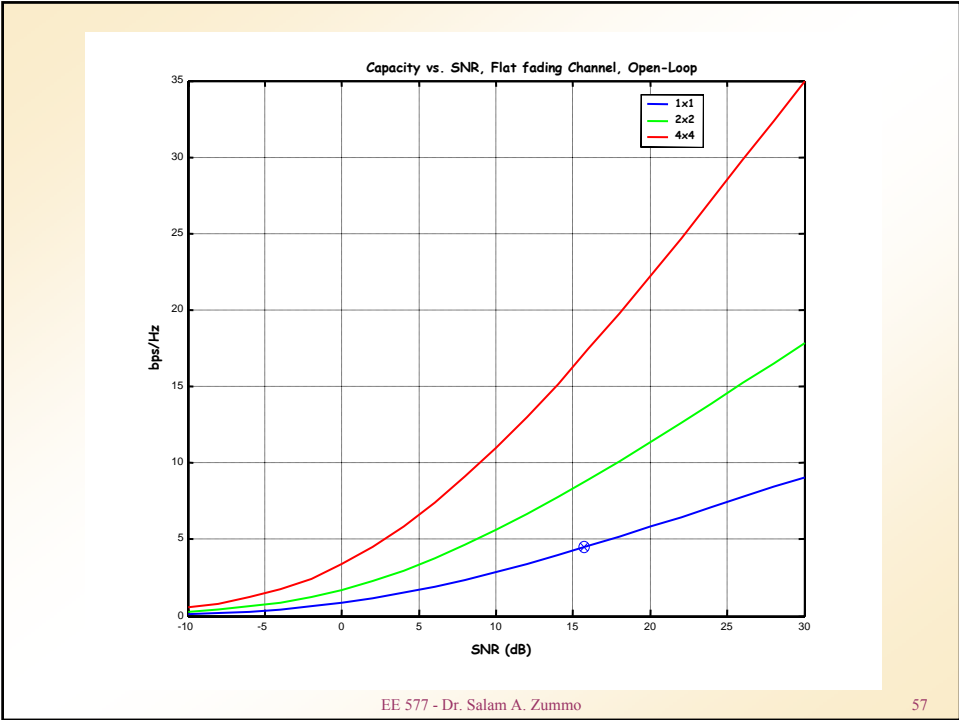
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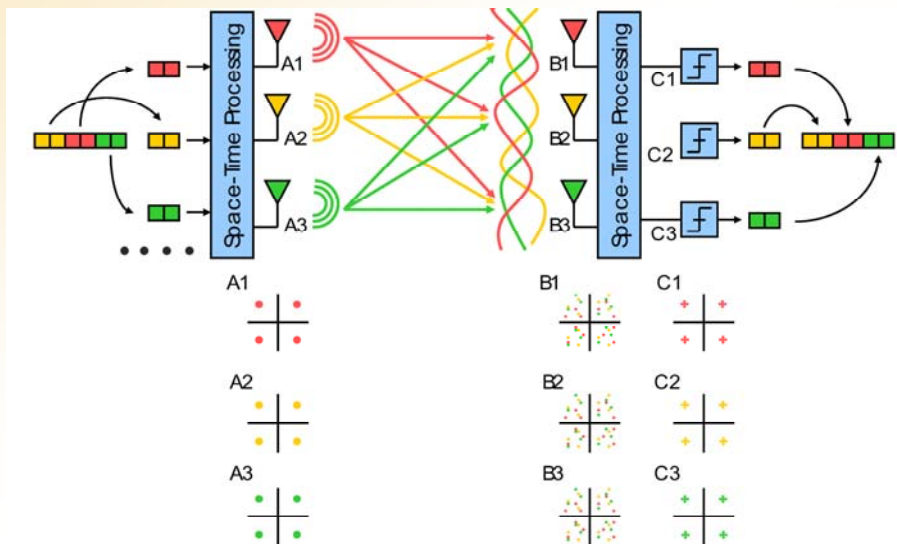


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Space-Time Coding



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Advantages of MIMO

- ❑ Order of magnitude increased rate, range and robustness
- ❑ Good cost-performance trade-off and scalability
- ❑ Same bandwidth and higher rates
 - => more efficient use of spectrum
- ❑ Increase downlink capacity
- ❑ Combats multipath fading
- ❑ Initial application in 3G WCDMA standard and 802.16 broadband wireless

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MIMO Techniques

❑ Open-Loop MIMO:

- ❑ Multiple coded data streams across multiple parallel transmitters
- ❑ Achieve diversity gain by space-time coding and/or interference cancellation at the receiver
- ❑ Linear increase in data rates with number of antennas

❑ Closed-Loop MIMO:

- ❑ Waterfilling to achieve higher data rates
- ❑ Transmitter requires channel knowledge
- ❑ Data rates **and** range/throughput gains

Adaptive Modulation

❑ Basic idea:

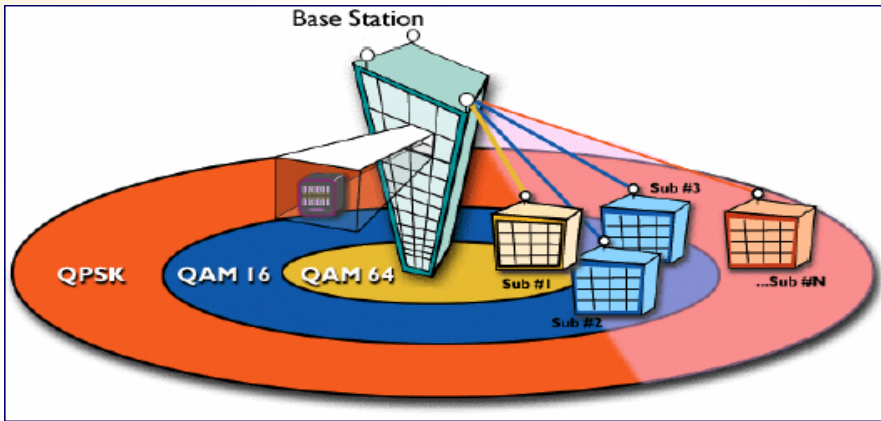
- ❑ Measure the channel at the receiver
- ❑ Feed the measurement back to the transmitter
- ❑ Adapt the transmission scheme relative to the channel estimate to **maximize the data rate, minimize transmit power or minimize BER**

❑ What to adapt?

- ❑ Constellation size/power
- ❑ Symbol time
- ❑ Coding rate/scheme

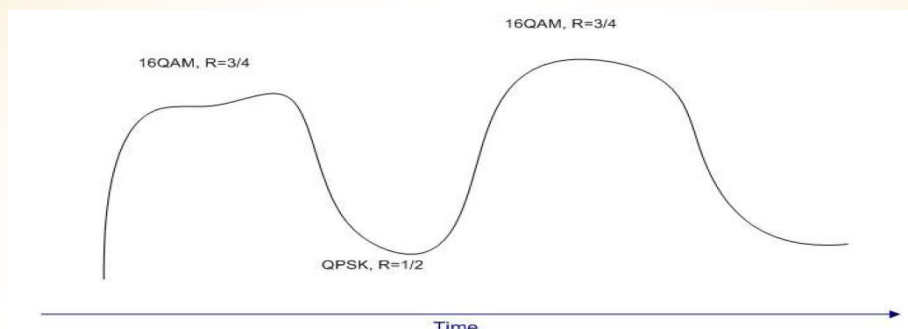
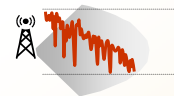
Bit Rates in IEEE 802.16a

- ❑ Bit rate shifting is achieved using adaptive modulation.
- ❑ When you are near to the BS => offered high speed,
- ❑ When you are far, reliability decreases => offered lower speed.



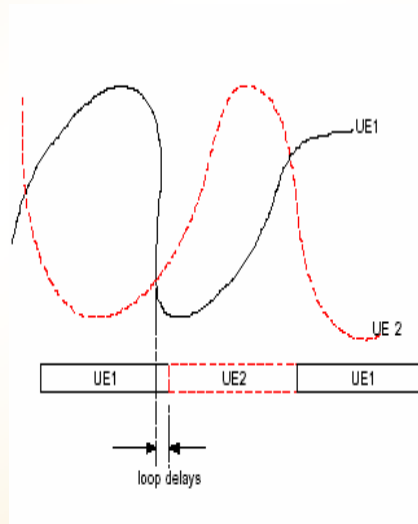
HSDPA Features

- ❑ Adaptive Modulation and Coding
 - ❑ Data rate adapted to radio conditions
 - ❑ 2 ms time basis



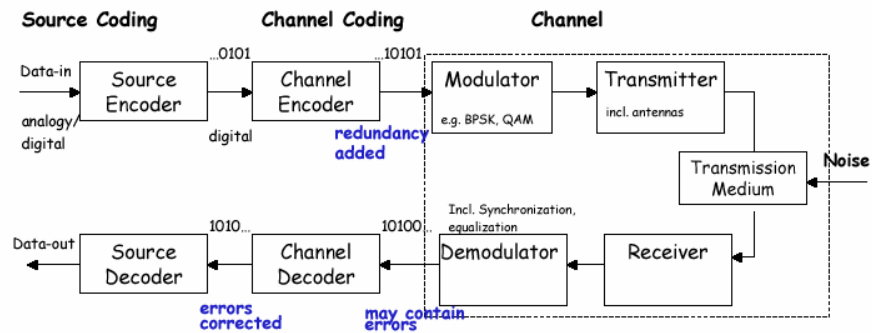
Multi-User Diversity

- ❑ Fast Scheduler
 - ❑ 2 ms time basis
 - ❑ Round Robin, Proportional Fair or Max-C/I
- ❑ Since users are independent with each other, let the users with good channel condition send at any given time → Multiuser Diversity.
- ❑ Fairness is also an important attribute



Channel Coding

Communication System



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Basic Channel Coding Concepts

Example: Binary Repetition Codes

- ❑ (3,1) code: 0 => 000, 1 => 111
- ❑ (3,1) repetition code can correct single errors
- ❑ Block error probability:

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

- ❑ **Gain:** For a BSC with $p = 10^{-2}$, $P_E = 3 \times 10^{-4}$.
- ❑ **Cost:** Expansion in bandwidth by 3 times

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Error Control Coding

- ❑ The function of the encoder is to introduce redundancy in the binary information sequence.
- ❑ Such redundancy is used in the receiver to overcome the effects of noise, interference and (fading) encountered through the channel.
- ❑ **Encoding** is the process of mapping k -bit information into a unique n -bit sequence called the “*codeword*”
- ❑ The *code rate* is defined as $R = k/n$

Shannon's Channel Capacity

- ❑ Shannon derived the capacity formula in 1948:

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

- ❑ W is the bandwidth in Hz
- ❑ S is the signal power in watts
- ❑ N is the total noise power
- ❑ The bandwidth efficiency can be found as:

$$\eta = \frac{\text{Transmission Rate}}{\text{Signal Bandwidth } W} \quad [\text{bits/s/Hz}]$$

$$\eta_{\max} = \log_2 \left(1 + \frac{S}{N} \right) \quad [\text{bits/s/Hz}]$$

Shannon's Channel Capacity

- The average signal power: $S = \frac{kE_b}{T} = RE_b$
- $$\eta_{\max} = \log_2 \left(1 + \frac{RE_b}{N_0W} \right)$$
- E_b is energy per bit
 - k is the number of bits transmitted per symbol
 - T is the duration of a symbol
 - $R = k/T$ is the transmission rate in bits/s
 - $N = N_0W$ is the total noise power
 - N_0 is the one-sided noise power spectral density

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Shannon's Channel Capacity

- The minimum bit energy required for reliable transmission (Shannon bound):

$$\frac{E_b}{N_0} \geq \frac{2^{\eta_{\max}} - 1}{\eta_{\max}}$$

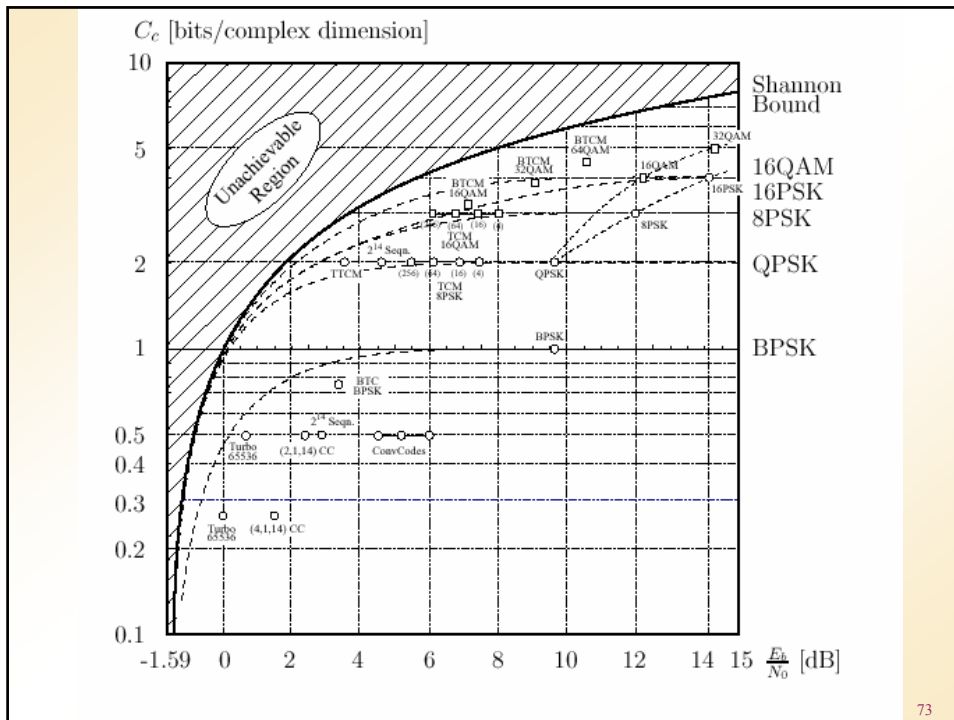
- In the case of infinite bandwidth, i.e., $\eta_{\max} \rightarrow 0$,

$$\frac{E_b}{N_0} \geq \lim_{\eta_{\max} \rightarrow 0} \frac{2^{\eta_{\max}} - 1}{\eta_{\max}} = \ln(2) = -1.59 \text{ dB}$$

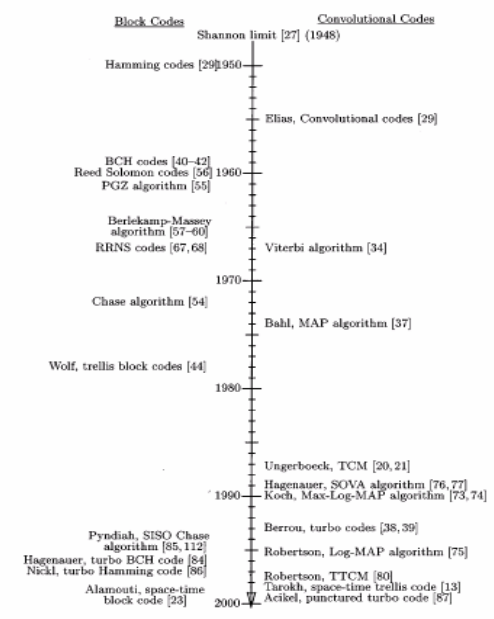
- This is the **minimum signal-to-noise ratio** required to reliably transmit one bit of information

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History of Channel Coding



Hamming Distance

- The **Hamming distance** between two codewords c_i and c_j , denoted by $d_H(c_i, c_j)$, is the number of elements at which they differ

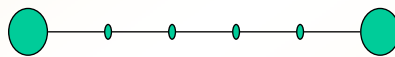
- **Examples:**

$$d_H(011,000) = 2$$

$$d_H(011,111) = 1$$

Error Correction and Detection

- Consider a code consisting of two codewords with Hamming distance d . How many errors can be detected? Corrected?



- # of errors that can be detected = $\lambda = d-1$

- # of errors that can be corrected = $t = \left\lfloor \frac{d-1}{2} \right\rfloor$

- In other words, for t -error correction: $d = 2t + 1$

Minimum Distance of a Code

- **Def.:** The **minimum distance** of a code C is the minimum Hamming distance between any two different codewords.

$$d_{\min} = \min_{i \neq j} d(c_i, c_j), \quad \forall c_i, c_j \in C$$

- A code with minimum distance d_{\min} can correct all error patterns up to and including t -error patterns, where

$$d_{\min} = 2t + 1$$

- It may be able to correct some higher weight error patterns, but not all.

Linear Block Codes

- A binary information vector (\mathbf{X}) of k bits is mapped onto a binary vector \mathbf{C} with $n > k$ bits
- The transformation is defined by a generator matrix \mathbf{G} which is $k \times n$ matrix
- The message is segmented into blocks of k bits
- There are 2^k codewords (one for each 2^k possible information vectors)
- A binary block code is **linear** if and only if the modulo-2 sum of two codewords is also a codeword

Generator Matrix

- ❑ Message vector: $\mathbf{X} = [x_{m1} \ x_{m2} \ x_{m3} \ \dots \ x_{mk}]$
- ❑ Codeword vector: $\mathbf{C} = [c_{m1} \ c_{m2} \ c_{m3} \ \dots \ c_{mn}]$
- ❑ Generator matrix of the code:

$$\mathbf{G} = \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{g}_2 \\ \vdots \\ \mathbf{g}_k \end{bmatrix} = \begin{bmatrix} g_{11} & g_{12} & \dots & g_{1n} \\ g_{21} & g_{22} & \dots & g_{2n} \\ \vdots & \vdots & \dots & \vdots \\ g_{k1} & g_{k2} & \dots & g_{kn} \end{bmatrix}$$

- ❑ Encoding is performed by using:

$$\mathbf{C}_m = \mathbf{X}_m \mathbf{G}$$

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(6,3) Linear Block Codes Example

Messages	Codewords
000	000000
100	110100
010	011010
110	101110
001	101001
101	011101
011	110011
111	000111

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Systematic Property

- A codeword is divided into two parts:
 - Message (systematic) bits (k)
 - Parity-check bits ($n-k$)
- A linear block code with this structure is referred to as a linear **systematic block code**
- The generator matrix for such a code is given by:

$$G = [I_k | P] = \begin{bmatrix} 1 & 0 & \dots & 0 & p_{11} & p_{12} & p_{13} & \dots & p_{1n-k} \\ 0 & 1 & \dots & 0 & p_{21} & p_{22} & p_{23} & \dots & p_{2n-k} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \dots & \vdots \\ 0 & 0 & \dots & 1 & p_{k1} & p_{k2} & p_{k3} & \dots & p_{kn-k} \end{bmatrix}$$

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Encoding Example

- The (7,4) linear code has the following matrix as generator matrix

$$G = \begin{bmatrix} \mathbf{g}_1 \\ \mathbf{g}_2 \\ \mathbf{g}_3 \\ \mathbf{g}_4 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 1 \end{bmatrix}$$

- If the message $\mathbf{x} = (1101)$, its corresponding codeword \mathbf{C} is given by:

$$\mathbf{C} = \mathbf{X} \cdot \mathbf{G}$$

$$= 1 \times \mathbf{g}_1 + 1 \times \mathbf{g}_2 + 0 \times \mathbf{g}_3 + 1 \times \mathbf{g}_4$$

$$= (1101000) + (0110100) + (1010001) = (0001101)$$

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Encoding Example

- The following matrix G is in the systematic form:

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

- The codeword C for a message X is given by:

$$C = X \cdot G$$

$$c_1 = x_1$$

$$c_2 = x_2$$

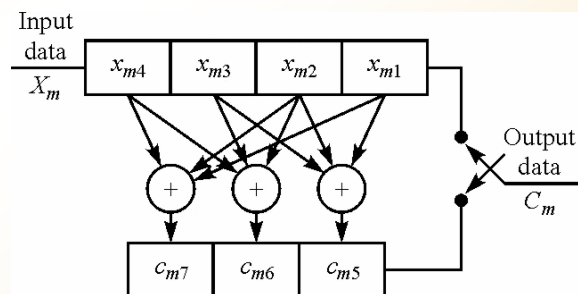
$$c_3 = x_3$$

$$c_4 = x_4$$

$$c_5 = x_1 + x_2 + x_3$$

$$c_6 = x_2 + x_3 + x_4$$

$$c_7 = x_1 + x_2 + x_4$$



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Parity-Check Matrix

- An $(n-k) \times n$ **parity check matrix** H has its rows orthogonal to all codewords generated by G
- Thus, a vector C is a codeword in the code generated by G if and only if $C \times H^T = \mathbf{0}$
- The matrix H is called a **parity check matrix**
- For a generator matrix $G = [I_k \ P] \Rightarrow H = [P^T \ I_{n-k}]$
- For the last example:

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

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Hamming Weight & Distance

- ❑ The weight of the codeword is the number of non-zero element it contains
- ❑ **Hamming distance** d_{ij} is a measure of the difference between C_i and C_j in any (n,k) block code
- ❑ The smallest value of the hamming weight is called the “**minimum distance**” d_{\min}
- ❑ The error detection capability of the code is:

$$d_{\min} - 1$$

- ❑ The error correction capability of the code is:

$$\left\lfloor \frac{1}{2}(d_{\min} - 1) \right\rfloor$$

Example: (7,4) Hamming Code

No.	Message	Codeword	No.	Message	Codeword
0	0000	0000000	8	0001	1010001
1	1000	1101000	9	1001	0111001
2	0100	0110100	10	0101	1100101
3	1100	1011100	11	1101	0001101
4	0010	1110010	12	0011	0100011
5	1010	0011010	13	1011	1001011
6	0110	1000110	14	0111	0010111
7	1110	0101110	15	1111	1111111

Linear Block Codes

❑ **Linear code:** The sum of any two codewords is a codeword.

❑ **Hamming codes** constitute a class of single-error correcting codes defined as:

$$n = 2^m - 1, k = n - m, m > 2$$

❑ The **minimum distance** of the code $d_{\min} = 3$

❑ Hamming codes are perfect codes.

CYCLIC CODES

❑ An (n,k) linear code \mathbf{C} is cyclic if every cyclic shift of a codeword in \mathbf{C} is also a codeword in \mathbf{C}

❑ If $c_0 \ c_1 \ c_2 \ \dots \ c_{n-2} \ c_{n-1}$ is a codeword, then the following sequences:

$$c_{n-1} \ c_0 \ c_1 \ \dots \ c_{n-3} \ c_{n-2}$$

$$c_{n-2} \ c_{n-1} \ c_0 \ \dots \ c_{n-4} \ c_{n-3}$$

$$\vdots \quad \vdots \quad \vdots \quad \quad \quad \vdots \quad \vdots$$

$$c_1 \ c_2 \ c_3 \ \dots \ c_{n-1} \ c_0$$

are all valid codewords.

Example

- The (7,4) Hamming code discussed before is cyclic:

1010001	1110010	0000000	1111111
1101000	0111001		
0110100	1011100		
0011010	0101110		
0001101	0010111		
1000110	1001011		
0100011	1100101		

Code Polynomial

- Let $\mathbf{c} = c_0 c_1 c_2 \dots c_{n-1}$

- The code polynomial of \mathbf{c} is:

$$c(X) = c_0 + c_1X + c_2X^2 + \dots + c_{n-1}X^{n-1}$$

where the power of X corresponds to the bit position, and the coefficients are **0**'s and **1**'s.

- **Example:**

1010001 $1+X^2+X^6$

0101110 $X+X^3+X^4+X^5$

Generator Polynomial

- ❑ All code polynomials are generated from one polynomial, the *generator polynomial*, using

$$c(X) = a(X)g(X)$$

- ❑ The *generator polynomial* completely defines the code
- ❑ The (7,4) Hamming code can be generated from the generator polynomial $1+X+X^3$

BCH Codes

- ❑ **Definition of BCH codes:**

For any positive integers m ($m > 2$) and t_0 ($t_0 < n/2$), there is a BCH binary code of length $n = 2^m - 1$ which corrects all combinations of t_0 or fewer errors and has no more than mt_0 parity-check bits.

Codeword length	$2^m - 1$
Number of parity-check bits	$n - k \leq mt_0$
Minimum distance	$d_{min} \geq 2t_0 + 1$

Table of Some BCH Codes

n	k	d (designed)	d (actual)	$g(X)^*$
7	4	3	3	13
15	11	3	3	23
15	7	5	5	721
15	5	7	7	2463
31	26	3	3	45
31	16	5	7	107657
31	11	7	11	5423325

* Octal representation with highest order at the left.
721 is 111 010 001 representing $1+X^4+X^6+X^7+X^8$

Reed-Solomon (RS) Codes

- A class of non-binary BCH codes.
- The codeword consists of n m -bit symbols.
- The parameters of the code are related by: $n-k = 2t$
- Example:** $m = 4, n = 15, k = 11,$
 - Codeword length is 15 symbols, or $15 \cdot 4 = 60$ bits.
 - It is a double-error correcting code ($t=2$).
 - It can correct any burst of 8 or less bit errors.

Convolutional Codes

- ❑ Convolutional codes differ from block codes in:
 - ❑ The encoder has memory
 - ❑ The n -bit output codeword depends on the k -bit input message and the previous input bits
- ❑ Convolutional coding is suitable for long messages such as streaming data (e.g., voice)
- ❑ The encoder consists of linear finite-state shift registers of K stages,
- ❑ Input bits are shifted k at a time to give n coded bits
- ❑ K is called the constraint length of the code.
- ❑ The rate of the code is $R = k/n$.

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Convolutional Code Example 1

- ❑ This code has $K = 3$, $k = 1$, $n = 3$, \Rightarrow rate = $1/3$
- ❑ The generator functions for the code are:

$$\mathbf{g}_1 = [100], \mathbf{g}_2 = [101], \mathbf{g}_3 = [111].$$
- ❑ The generator functions can be represented either in **octal form** or by **generator polynomials**:

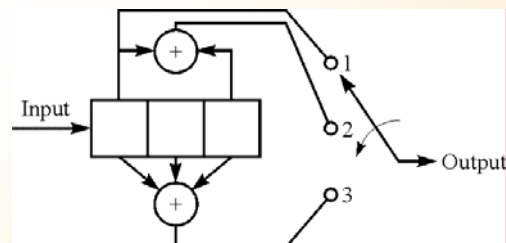
❑ Octal form: $(4, 5, 7)_8$

❑ Generator polynomials:

$$g_1(x) = 1$$

$$g_2(x) = 1 + x^2$$

$$g_3(x) = 1 + x + x^2$$



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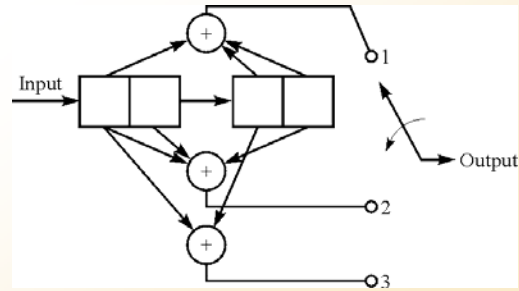
Convolutional Code Example 2

- This code has $K = 2$, $k = 2$, $n = 3$, and a rate of $2/3$
- The generator functions for this code are $g_1 = [1011]$, $g_2 = [1101]$, $g_3 = [1010]$ and in octal form, these generators are: $(13, 15, 12)_8$.
- Generator polynomials can be written as:

$$g_1(x) = 1 + x^2 + x^3$$

$$g_2(x) = 1 + x + x^3$$

$$g_3(x) = 1 + x^2$$



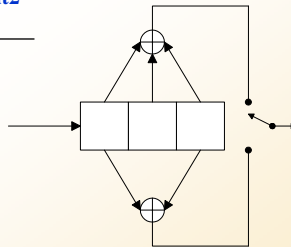
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Encoding Example

- Find the state changes and the resulting output codeword sequence for the message $m = 1 1 0 1 1 0 0$. Assume that the initial contents of the encoder are all zero

Input bit	Register Content	State at time t_i	State at time t_{i+1}	output1	Output2
-	000	00	00	-	-
1	100	00	10	1	1
1	110	10	11	0	1
0	011	11	01	0	1
1	101	01	10	0	0
1	110	10	11	0	1
0	011	11	01	0	1
0	001	01	00	1	1



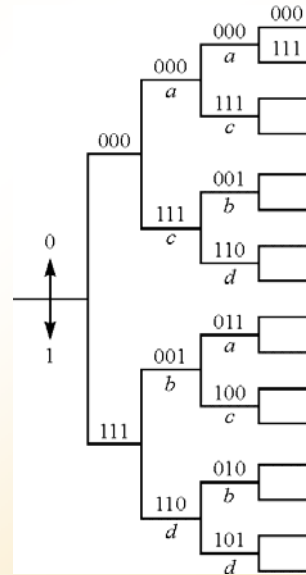
- Output sequence: $C = 11 01 01 00 01 01 11$

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Convolutional Codes Representation

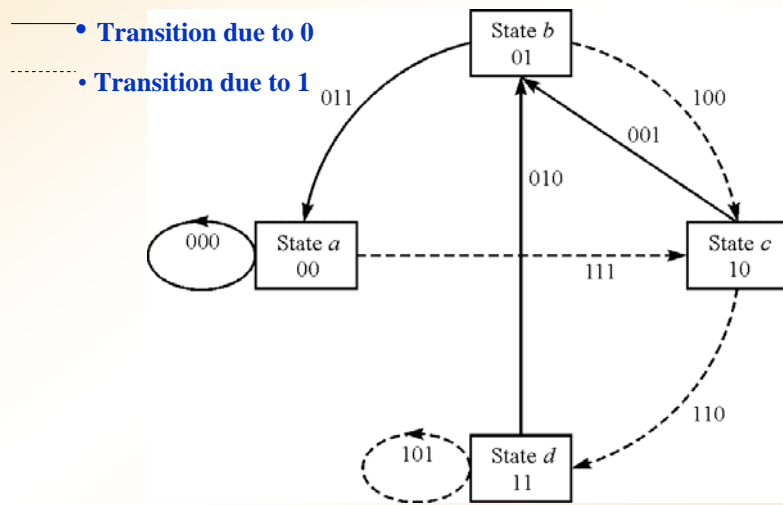
- ❑ Three methods to represent a convolutional code:
 - ❑ Tree diagram
 - ❑ Trellis diagram
 - ❑ State diagram
- ❑ A tree diagram for the rate-1/3, $K=3$ code is shown



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State Diagram

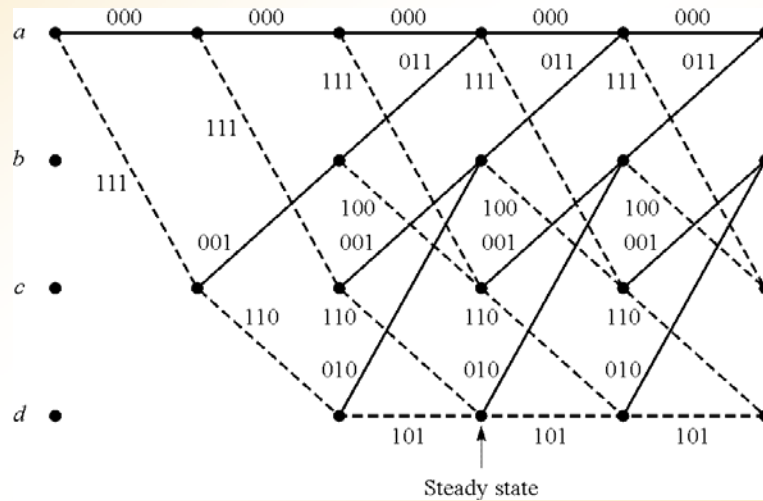
A state diagram for the rate-1/3, $K=3$ code is shown:



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Trellis Diagram

A trellis diagram for the rate-1/3, $K=3$ code is shown:



Decoding of Convolutional Codes

❑ Hard Decoding:

The received symbols at the output of the demodulator are quantized into two levels; zero and one, and fed to the decoder

❑ Soft Decoding:

- ❑ The received symbols at the output of the demodulator are unquantized value (analog value) is used and fed to the decoder
- ❑ Yields a gain of 2 - 2.2 dB compared to hard decoding

Decoding of Convolutional Codes

Maximum Likelihood Sequence Decoding:

- ❑ A convolutional code is converted into a block code of length L by feeding zeros at the end of the input message to force the encoder back to the zero state
- ❑ Find the codeword with closest Hamming distance (hard decoding) or Euclidean distance (soft decoding) from the possible 2^L codewords.

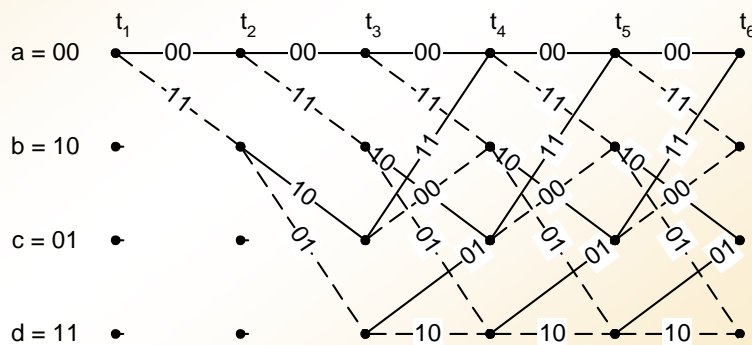
Viterbi Algorithm:

- ❑ Computes the distance between the received sequence and all the potential trellis paths
- ❑ At each stage, keeps one “most likely” (surviving) path for each state

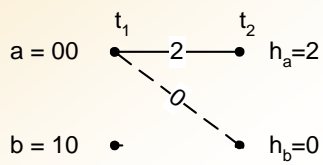
Viterbi Decoding Example

- ❑ For the convolutional encoder, assume that the received sequence $z = 11 \ 01 \ 01 \ 10 \ 01$.

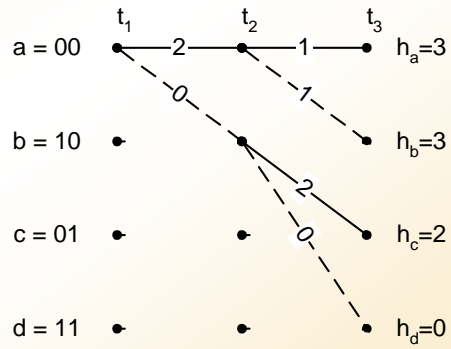
- ❑ Find the input message sequence m .



Viterbi Decoding Example



(a)

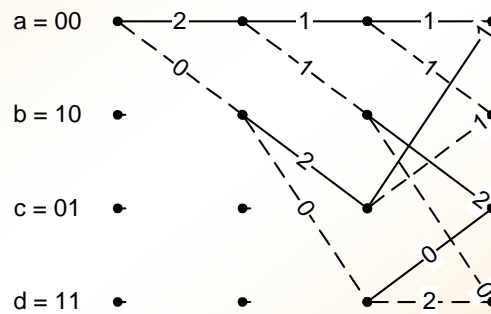


(b)

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Viterbi Decoding Example

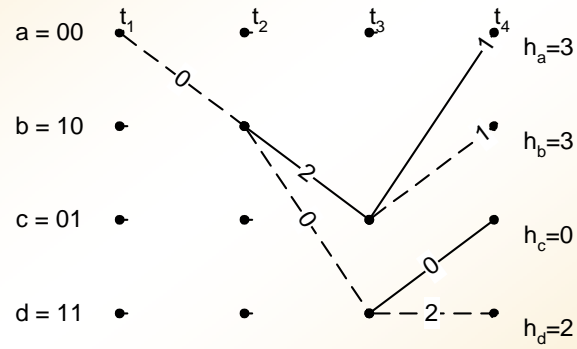


(c)

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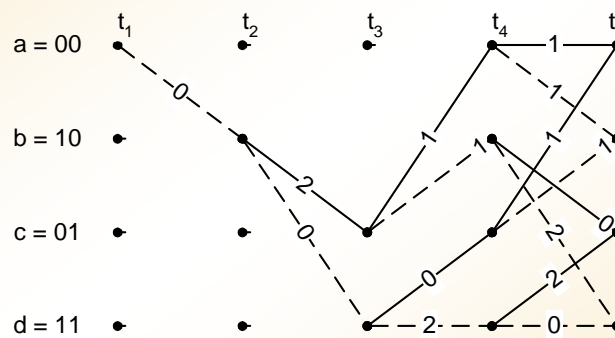
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Viterbi Decoding Example



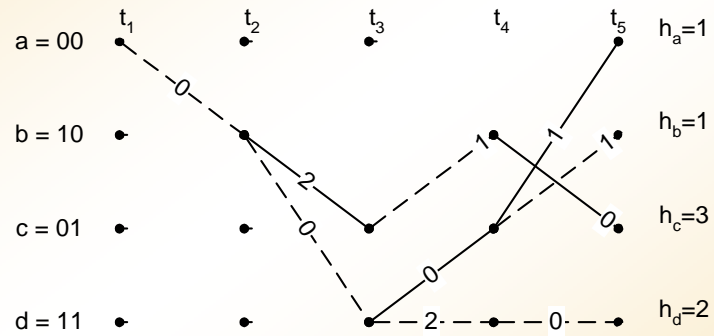
(d)

Viterbi Decoding Example



(e)

Viterbi Decoding Example

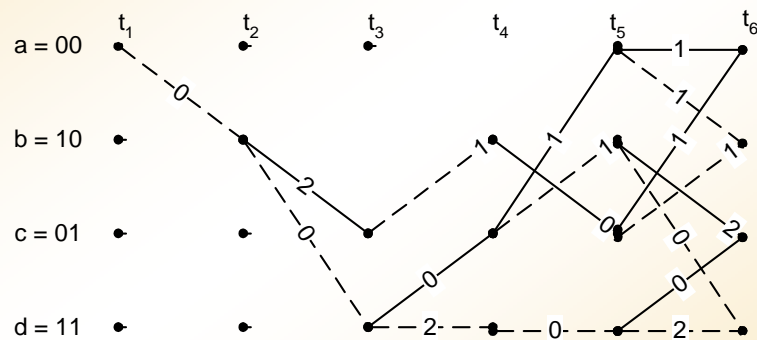


(f)

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Viterbi Decoding Example

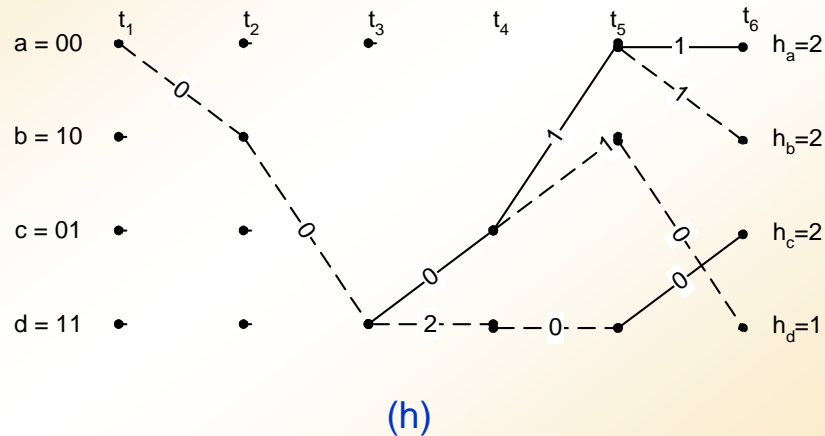


(g)

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Viterbi Decoding Example



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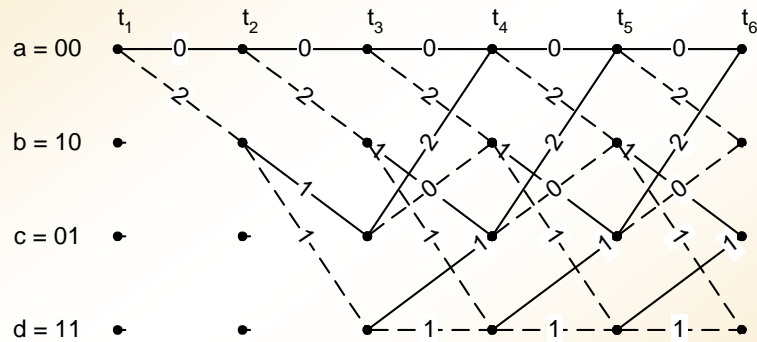
Distance Properties of Convolutional Codes

- Convolutional codes are linear codes
- The free distance of convolutional codes is associated with the path that starts and ends in the all zero state and does not return in between.
- So given the *all-zero* transmission an error occurs whenever the *all-zero* path does not survive.
- The minimum distance is found by exhaustively searching every path from the *all-zero* state to the *all-zero* state.

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Distance Properties of Convolutional Codes



Trellis diagram labelled with distance from the *all-zero* path for the encoder

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Distance Properties of Convolutional Codes

□ Examining the previous trellis diagram it is clear that the *free* distance of the code is 5

□ This means that this code can correct up to:

$$\left\lfloor \frac{1}{2}(d_{\min} - 1) \right\rfloor = 2 \quad \text{errors}$$

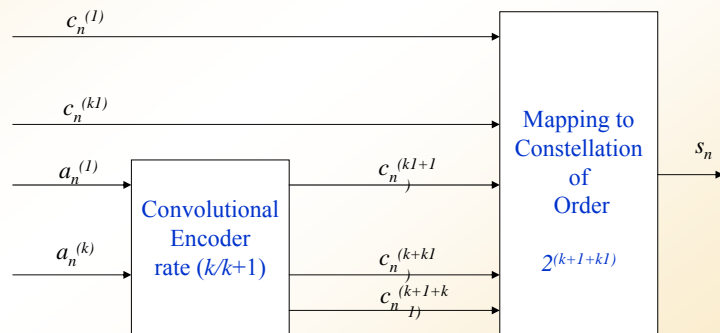
□ A more closed form expression can be obtained by finding the transfer function of the code

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Trellis Coded Modulation (TCM)

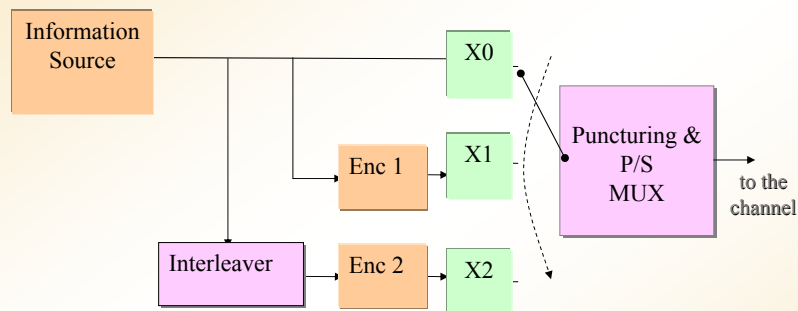
- ❑ Convolutional encoder with a signal output (instead of binary bits)
- ❑ Encoding is done to maximize some distance criterion in the signal constellation



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Turbo Codes



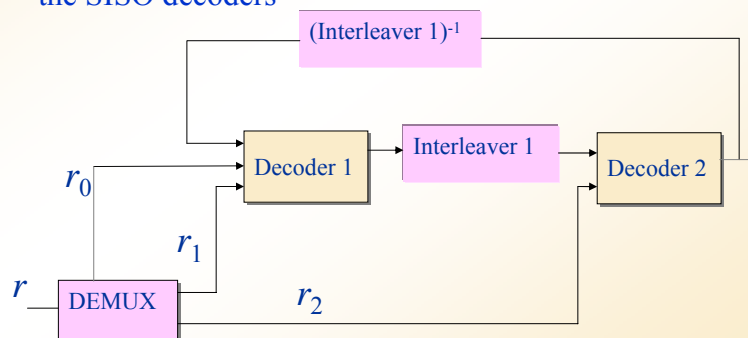
- ❑ Achieves performance close to channel capacity over AWGN and flat fading channels
- ❑ Information is encoded by two encoders after being interleaved

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Turbo Decoder

- ❑ **Iterative decoding** is composed of:
 - ❑ Two soft-input soft-output decoders for codes 1 and 2
 - ❑ Interleaver – DeInterleaver pair
 - ❑ Soft information about message bits are exchanged between the SISO decoders



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Coding and Interleaving

- ❑ Codes designed for the AWGN channel do not work well in fading channels due to burst errors
- ❑ This can be compensated for by using standard AWGN channel combined with an interleaver to spread burst errors at the decoder

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Coding and Interleaving

- Channel coding is a form of time diversity
- Independent fading is needed on each bit in a codeword to get the diversity gain
- Interleaving breaks the memory of the channel and provides independent fading for each bit
- The cost of interleaving is increased complexity and delay