無線通訊系統 (Wireless Communications Systems)

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Prof. Tsai

Chapter 3 Physical Layer Technologies for Wireless Communication Systems

Modulation Schemes

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Modulation

- Message information can be transmitted in the amplitude,
 frequency or phase of the carrier
- For mobile radio applications:
 - We shall use bandwidth and power resources most efficiently
- **Bandwidth efficiency:** measured in bits/sec/Hz
 - The modulation must have compact power density spectrum
- Good BER performance:
 - Good BER must be maintained in mobile radio environments
 - The modulation must be able to prevail over fading, ISI, Doppler spread, adjacent and co-channel interference, and thermal noise
- Envelope properties:
 - Portable transmitters normally use nonlinear power amplifiers
 - Modulation with a **relatively constant envelope** shall be used

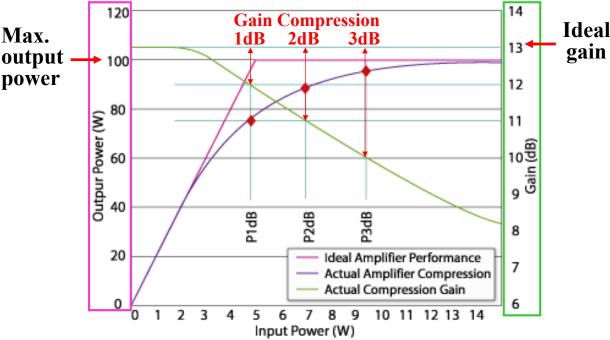
Power Amplifier

- A RF **power amplifier** (PA) can amplify a small input RF signal of a certain shape to a larger one with the same shape at the output.
- The gain is defined as the ratio of output power to input power.
- Since the maximum output power is limited, a PA will definitely be **saturated** when input power gradually increases.
- The saturation of a PA is defined as the situation that the output signal **is not proportional to** the input signal.
 - **Distortion** occurs in the output signal
- It is highly desirable to operate a PA in the linear region
 - The input signal is faithfully re-produced at the output

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Power Amplifier

Theoretical and actual power-amplifier gain



Power Amplifier

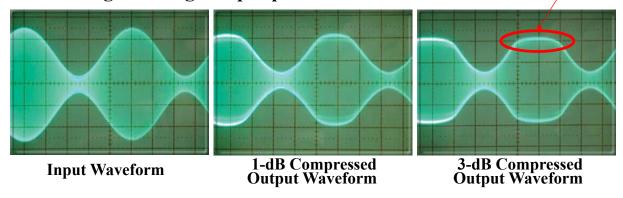
- The amount of usable power depends on the applications
 - It depends on how much distortion can be tolerated
- A commonly used figure-of-merit is the **1-dB compression point**. This is the output power level at which the gain is compressed by 1 dB.
 - However, the decreased gain affects only those parts of the signal with amplitudes close to or greater than the 1-dB compression point.

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Power Amplifier

- For 1-dB compressed output waveform, the maximum input power is equal to the "P1dB" input power
 - Smaller average output power with less distortion
- For 3-dB compressed output waveform, the maximum input power is equal to the "P3dB" input power

 High frequency components exist
 - Larger average output power with more distortion



Power Amplifier

- We may select the required amplifier power by **backing off** an acceptable amount from P1dB to prevent signal distortion.
- For example, if an application requires a maximum output power of 50 W, the system designer may select a PA with a P1dB of 100 W or higher to ensure that the PA is within its linear region of operation.
 - It depends upon the **linearity requirement** of the system.
- For the same tolerable distortion, a signal with a high peak-to-average ratio will **degrade** the **average output signal power**, as well as the efficiency of the PA.
 - Or equivalently, it increases the linearity and dynamic range requirements of the PA

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Modulation

- First generation systems:
 - FM modulation (AMPS, TACS, NMT)
- Second generation systems:
 - $-\pi/4$ -DQPSK (IS-136, PDC, PHS)
 - **GMSK** (GSM, DCS1800, DECT)
- Third generation systems:
 - CDMA QPSK/Offset QPSK (cdma2000, WCDMA)
- Fourth generation systems:
 - OFDM **BPSK, QPSK, 16-QAM, 64-QAM** (WiMAX, LTE)

Phase Shift Keying (PSK)

• The *M*-PSK complex envelope can be expressed as

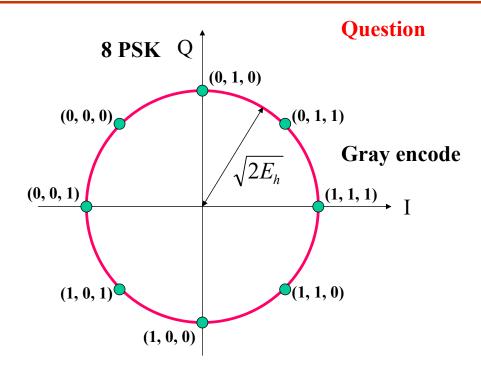
$$\tilde{s}(t) = A \sum_{n} b(t - nT, \mathbf{x}_{n})$$
$$b(t, \mathbf{x}_{n}) = h_{a}(t) \exp(j\theta_{n})$$

- $-\theta_n = \frac{2\pi}{M}x_n + \theta_0$ and θ_0 is an arbitrary constant phase
- $-x_n = n, n \in \{0, 1, ..., M-1\}$
- *M* is the alphabet size
- $-h_a(t)$ is the amplitude shaping pulse
 - Normally chosen to be a raised cosine function
- The complex envelope is

$$\tilde{s}_n(t) = Ah_a(t)e^{j\theta_n}, \quad n = 0, 1, \dots, M-1$$

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Phase Shift Keying (PSK)

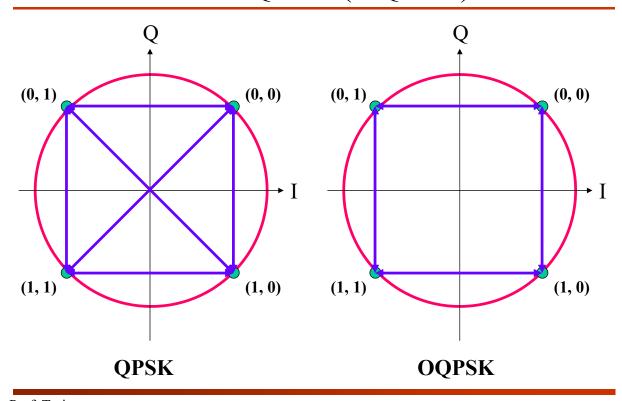


Offset QPSK (OQPSK)

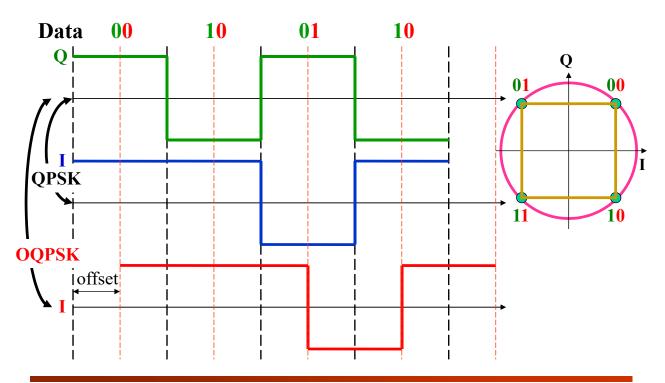
- The QPSK signal can have either ± 90° or ± 180° phase shift from one baud interval to the next
- The OQPSK signal can change by only $\pm 90^{\circ}$
- The phase trajectory does not pass through the origin
 - Reduce the peak-to-average ratio of the complex envelope
 - Make the OQPSK signal less sensitive to amplifier nonlinearity than the QPSK signal

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Offset QPSK (OQPSK)



Offset QPSK (OQPSK)



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$\pi/4$ -Differential QPSK ($\pi/4$ -DQPSK)

- QPSK: 4 carrier phases (Coherent detection is essential)
- $\pi/4$ -DQPSK: 8 carrier phases (Non-coherent detection)
- For quaternary source sequence $\{x_n\}, x_n \in \{\pm 1, \pm 3\}$

$$\Delta \theta_n = \theta_n - \theta_{n-1} = \begin{cases} -3\pi/4, & x_n = -3\\ -\pi/4, & x_n = -1\\ +\pi/4, & x_n = +1\\ +3\pi/4, & x_n = +3 \end{cases}$$

– The phase differences must be $\pm \pi/4$ and $\pm 3\pi/4$

$$\Delta \theta_n = x_n \frac{\pi}{4}$$

• The complex envelope is expressed as

$$\widetilde{s}(t) = A \sum_{n} b(t - nT, \mathbf{x}_n)$$

$\pi/4$ -Differential QPSK ($\pi/4$ -DQPSK)

- where

$$b(t, \mathbf{x}_n) = h_a(t) \exp\left\{ j \left(\theta_{n-1} + x_n \frac{\pi}{4} + \theta_0 \right) \right\}$$
$$= h_a(t) \exp\left\{ j \frac{\pi}{4} \left(\sum_{k=-\infty}^{n-1} x_k + x_n \right) + j \theta_0 \right\}$$

- Change the phase for every symbol: the symbol synchronization is much easier than QPSK
- Non-coherent detection can be applied in receivers
- There are two signal sets:

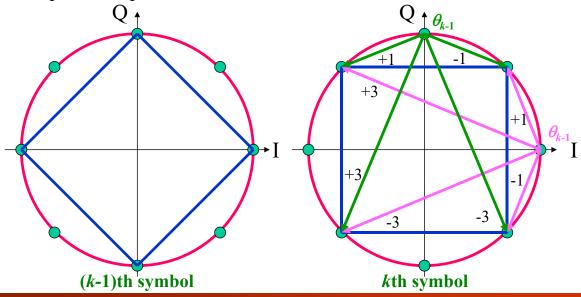
$$\{0, \pi/2, \pi, 3\pi/2\}$$

 $\{\pi/4, 3\pi/4, 5\pi/4, 7\pi/4\}$

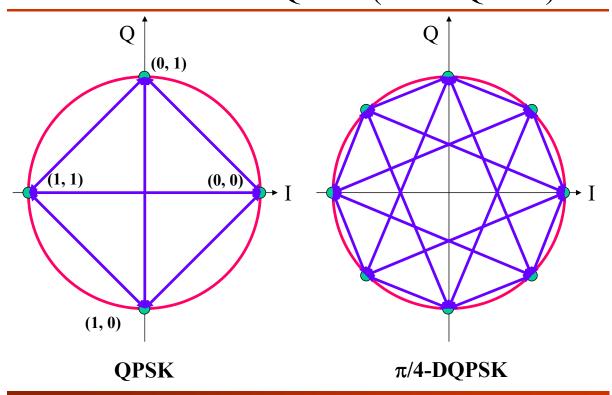
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$\pi/4$ -Differential QPSK ($\pi/4$ -DQPSK)

- The phase trajectory does not pass through the origin
 - Reduce the linearity and dynamic range requirements of the power amplifier

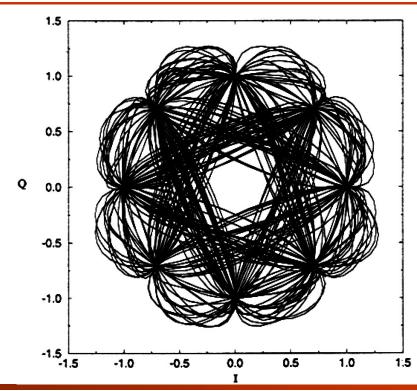


$\pi/4$ -Differential QPSK ($\pi/4$ -DQPSK)



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$\pi/4$ -Differential QPSK ($\pi/4$ -DQPSK)



Continuous Phase Modulation (CPM)

- Continuous Phase Modulation: the carrier phase varies in a continuous manner
- CPM is attractive because it has **constant envelope** and **excellent spectral characteristics**
- The complex envelope is

$$\tilde{s}(t) = A \exp\left\{j\left(\phi(t) + \theta_0\right)\right\}$$
$$\phi(t) = 2\pi \int_0^t \sum_{k=0}^\infty h_k x_k h_f(\tau - kT) d\tau$$

- $-\phi(t)$: the excess phase: the phase difference to the zero phase
- $\{x_k\}$: the data symbol sequence
- $\{h_k\}$: the sequence of modulation indices
- -h(t): the frequency shaping function

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Continuous Phase Modulation (CPM)

- $h_t(t)$ is zero for t < 0 and t > LT
 - Full response CPM: L = 1
 - Partial response CPM: L > 1
- The phase shaping function:

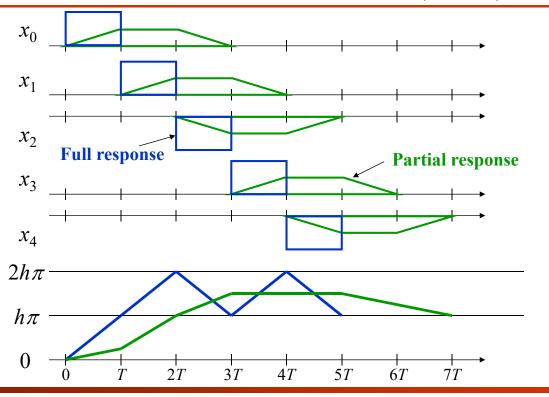
$$\beta(t) = \begin{cases} 0, & t < 0 \\ \int_0^t h_f(\tau) d\tau, & 0 \le t \le LT \\ 1/2, & t \ge LT \end{cases}$$

• For a full response CPM and $h_k = h$, the excess phase is

$$\phi(t) = 2\pi h \left[\int_{0}^{nT} \sum_{k=0}^{n-1} x_k h_f(\tau - kT) d\tau + \int_{nT}^{t} x_n h_f(\tau - nT) d\tau \right]$$

$$= \pi h \sum_{k=0}^{n-1} x_k + 2\pi h x_n \beta(t - nT) \text{ Response of the previous symbols} \text{ Response of the current symbol}$$

Continuous Phase Modulation (CPM)



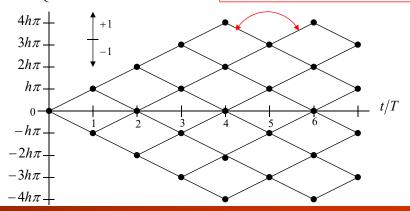
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Continuous Phase Frequency Shift Keying

- Continuous phase frequency shift keying (CPFSK):
 - A special type of full response CPM
 - By using the **rectangular** shaping function $h_t(t) = u_T(t)$

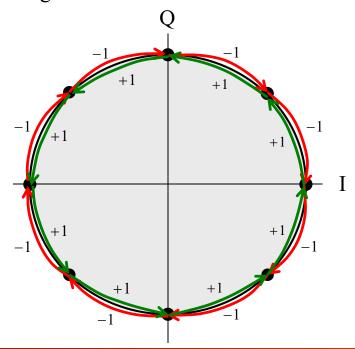
$$\beta(t) = \begin{cases} 0, & t < 0 \\ t/2T, & 0 \le t \le T \\ 1/2, & t \ge T \end{cases}$$
 Two slopes
$$\Rightarrow \text{Two frequencies}$$
 Representing '+1' & '-1'

• The phase tree:



Continuous Phase Frequency Shift Keying

• Binary CPM signal with h = 1/4



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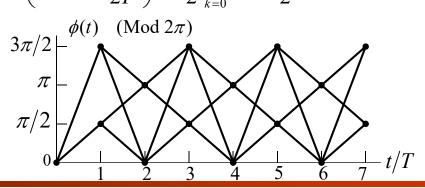
Minimum Shift Keying (MSK)

- Minimum Shift Keying is a special case of binary CPFSK
 - With modulation index h = 1/2
- Carrier phase: Zero phase

$$\phi_{c}(t) = 2\pi f_{c}t + \frac{\pi}{2} \sum_{k=0}^{n-1} x_{k} + \frac{\pi}{2} x_{n} \frac{t - nT}{T}$$

$$= \left(2\pi f_{c} + \frac{\pi x_{n}}{2T}\right)t + \frac{\pi}{2} \sum_{k=0}^{n-1} x_{k} - \frac{\pi n}{2} x_{n}$$

• Excess phase:



Minimum Shift Keying (MSK)

• The MSK band-pass waveform is

$$s(t) = A\cos\left(2\pi\left(f_c + \frac{x_n}{4T}\right)t + \frac{\pi}{2}\sum_{k=0}^{n-1}x_k - \frac{\pi n}{2}x_n\right), \quad nT \le t \le (n+1)T$$

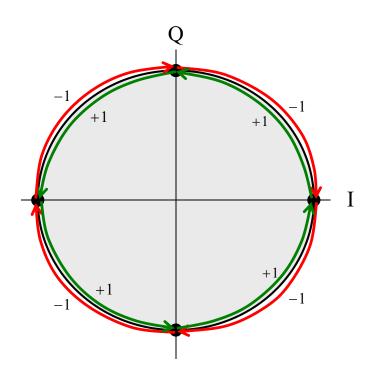
• Note that an MSK signal has one of two possible frequencies:

$$f_L = f_c - \frac{1}{4T}$$
, and $f_U = f_c + \frac{1}{4T}$

• The difference $\Delta f = f_U - f_L = \frac{1}{2T}$ is the minimum frequency separation to ensure **orthogonality** between two sinusoid of duration T with coherent demodulation

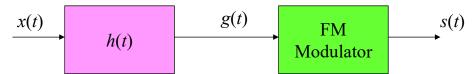
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Minimum Shift Keying (MSK)



Gaussian Minimum Shift Keying (GMSK)

- Gaussian Minimum Shift Keying is achieved by low-pass filtering the modulating signal prior to modulation
 - The filtering removes the higher frequency components
 - ⇒ It results in a more compact spectrum
 - ⇒ It introduces the inter-symbol interference (ISI)



• The transfer function of the pre-modulation filter (low-pass) is

$$H(f) = \exp\left\{-\left(\frac{f}{B}\right)^2 \frac{\ln 2}{2}\right\}$$

- where B is the 3 dB bandwidth of the filter

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Gaussian Minimum Shift Keying (GMSK)

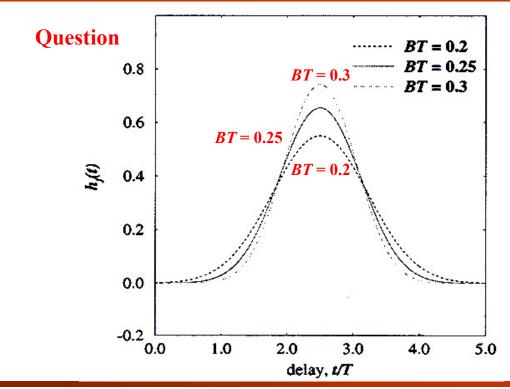
- The GMSK frequency shaping pulse filter:
 - The normalized filter bandwidth is BT
 - The shaping pulse duration > T ⇒ ISI
 - -BT decreases ⇒ more compact in power spectral density ⇒ ISI increases ⇒ the bit error rate performance is degraded
- In GSM systems:

$$x_n \in \{-1, +1\}$$

$$h = 1/2$$

$$BT = 0.3$$

Gaussian Minimum Shift Keying (GMSK)

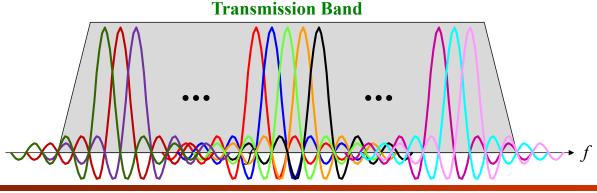


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OFDM

Orthogonal Frequency Division Multiplexing

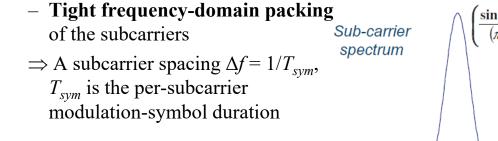
- Orthogonal frequency division multiplexing (OFDM) is a promising technique because of its
 - High bandwidth efficiency and
 - Resistance to multipath fading
- OFDM transmission can be regarded as a kind of multi-carrier transmission.



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Orthogonal Frequency Division Multiplexing

- However, the following characteristics distinguish OFDM from a straightforward multi-carrier extension:
 - The use of a typically very large number of relatively narrowband subcarriers (e.g., several hundred subcarriers)
 - Simple rectangular pulse shaping is used
 - \Rightarrow A sinc-square-shaped per-subcarrier spectrum



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 $-4 \Delta f - 3 \Delta f - 2 \Delta f - \Delta f$

 $2 \Delta f \quad 3 \Delta f$

Orthogonal Frequency Division Multiplexing

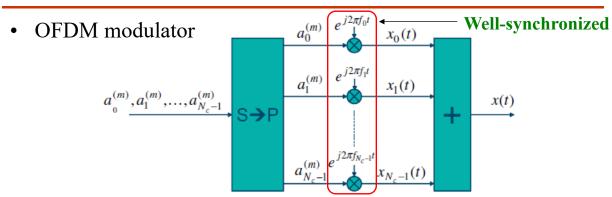
- OFDM can be regarded as a frequency-division multiplexing (FDM) scheme.
- A large number of closely-spaced **orthogonal subcarriers** are used to carry data.
- In complex baseband notation, a basic OFDM signal x(t) during the time interval $mT_{sym} \le t \le (m+1)T_{sym}$ is expressed as:

$$x(t) = \sum_{k=0}^{N_c - 1} x_k(t) = \sum_{k=0}^{N_c - 1} a_k^{(m)} e^{j2\pi k\Delta ft}$$

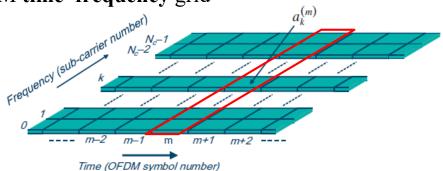
- $-x_k(t)$ is the *k*-th modulated subcarrier with $f_k = k\Delta f$
- $-a_k^{(m)}$ is the modulation symbol applied to the k-th subcarrier during the m-th OFDM symbol interval

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OFDM Modulation



• OFDM time-frequency grid



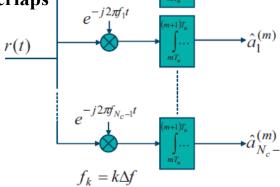
OFDM Demodulation

 The basic principle of OFDM demodulation consists of a bank of correlators, each of which corresponds to a subcarrier.

• Because of the orthogonality between subcarriers, in the ideal case, two OFDM subcarriers **do not** cause any interference to each other after demodulation.

 Despite that the spectrum of neighbor subcarriers overlaps

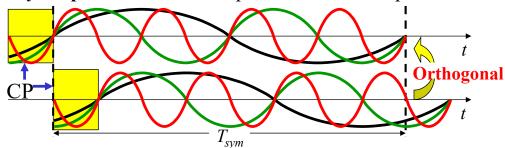
OFDM demodulator



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OFDM Orthogonality

- The **subcarrier orthogonality** is due to
 - The specific **frequency-domain structure** of each subcarrier
 - The specific choice of a subcarrier spacing Δf equal to the persubcarrier modulation-symbol rate $1/T_{sym}$
- Any **corruption** of the frequency-domain structure may lead to destruction in orthogonality ⇒ causing **mutual interference**
 - Time dispersion, **Doppler spread**, ...
 - Cyclic prefix is inserted to prevent ISI in multipath channels



OFDM Implementation

- OFDM allows for low-complexity implementation by means of computationally efficient Fast Fourier Transform (FFT) processing.
 - The **sampling rate** f_s is a multiple of the subcarrier spacing Δf

$$f_s = 1/T_s = N \times \Delta f$$

- The FFT size N should exceed N_c with a sufficient margin, $N > N_c$
- The time-discrete OFDM signal can be expressed as
 - The sequence x_n is the Inverse Discrete Fourier Transform (IDFT) of the symbols $a_0, a_1, ..., a_{Nc-1}$, extended with zeros to length N

$$x_{n} = x(nT_{s}) = \sum_{k=0}^{N_{c}-1} a_{k} e^{j2\pi k\Delta f nT_{s}}$$

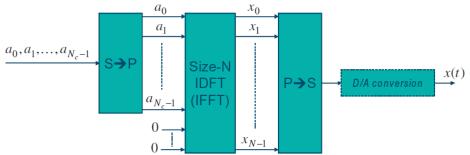
$$= \sum_{k=0}^{N_{c}-1} a_{k} e^{j2\pi kn/N} = \sum_{k=0}^{N_{c}-1} a'_{k} e^{j2\pi kn/N}$$

$$a'_{k} = \begin{cases} a_{k} & 0 \le k < N_{c} \\ 0 & N_{c} \le k < N \end{cases}$$

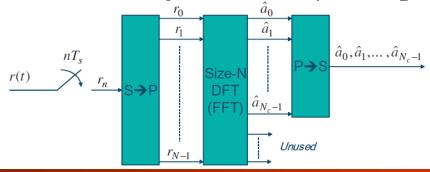
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OFDM Implementation

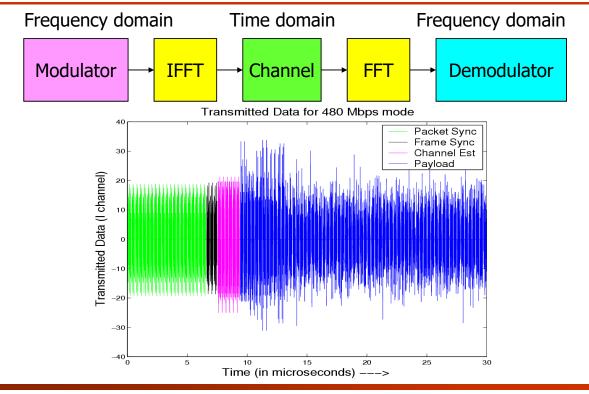
OFDM modulation by means of IFFT processing



OFDM demodulation by means of FFT processing



OFDM Transmission



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OFDM Transmission

Question

- The data are divided into several parallel data streams or channels, one for each subcarrier.
- Each subcarrier applies a **conventional modulation scheme** (PSK, QAM, ...) at a low symbol rate.
- OFDM may be regarded as using many **slowly-modulated narrowband signals** rather than one rapidly-modulated wideband signal.
 - The channel equalization is simplified
- OFDM requires very accurate **frequency synchronization** between the receiver and the transmitter
 - Including the **carrier** frequency and **sampling** frequency
 - Otherwise inter-carrier interference (ICI) is introduced

Digital Signaling on Flat Fading Channels

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BER-BPSK

The probability of bit error can be expressed as

$$P_b(\gamma_b) = Q(\sqrt{2\gamma_b})$$
$$\gamma_b = \alpha^2 E_b / N_0$$

• The pdf of energy-to-noise ratio for flat Rayleigh fading channels is

$$p_{\gamma_b}(x) = (1/\overline{\gamma}_b)e^{-x/\overline{\gamma}_b}$$

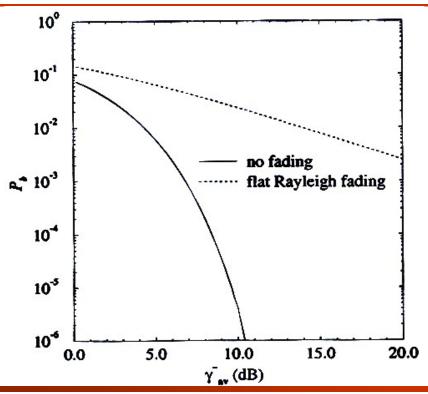
• The average probability of bit error becomes

$$P_b = \int_0^\infty P_b(x) p_{\gamma_b}(x) dx$$

$$= \int_0^\infty Q(\sqrt{2x}) p_{\gamma_b}(x) dx$$

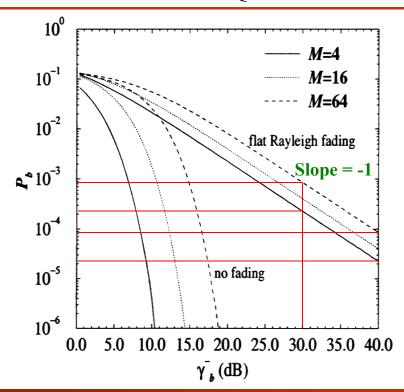
$$= \frac{1}{2} \left[1 - \sqrt{\frac{\overline{\gamma}_b}{1 + \overline{\gamma}_b}} \right] \approx \frac{1}{4\overline{\gamma}_b} \quad \text{for } \overline{\gamma}_b >> 1$$

BER-BPSK and QPSK



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BER-M-QAM

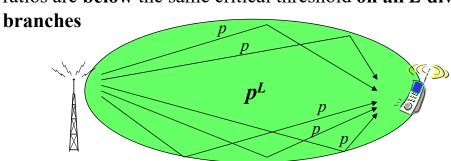


Diversity Techniques

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Diversity Techniques

- Diversity is a very effective solution for combating fading:
 - It can provide the receiver with multiple faded replicas of the same information signal
 - Let p denote the probability that the instantaneous signal-to-noise ratio is **below** some critical threshold **on a particular diversity branch**
 - $-p^L$ is the probability that all the instantaneous signal-to-noise ratios are **below** the same critical threshold **on all** L **diversity**



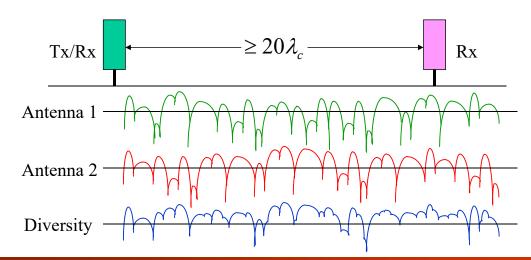
Diversity Techniques

- Diversity techniques:
 - Space
 - Angle
 - Polarization
 - Field
 - Frequency
 - Multipath
 - Time

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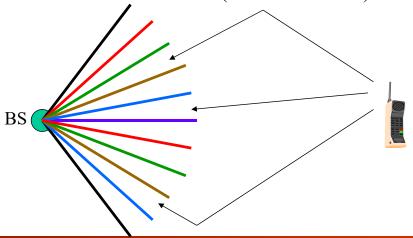
Space Diversity

- Space diversity (generally applied at BSs):
 - Achieved by using multiple receive antennas
 - The spatial separation between the antennas is chosen so that the diversity branches experience uncorrelated fading



Angle Diversity

- Angle diversity:
 - A number of directional antennas are required
 - Each antenna selects plane waves arriving from a narrow range of angles
 - All branches are uncorrelated (WSSUS channel)



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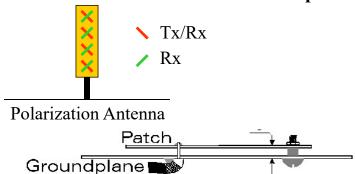
Polarization Diversity

• Polarization diversity:

- Question
- The scattering environment tends to **depolarize** a signal
- Receive antennas having different polarizations can be used to obtain diversity

At subscriber units: combination of monopole and patch antennas

At base stations: slant 45° monopole



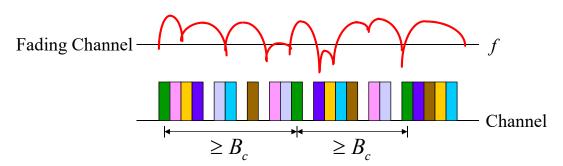
Field Diversity

- Field diversity:
 - The electric field and magnetic field components at any point are uncorrelated

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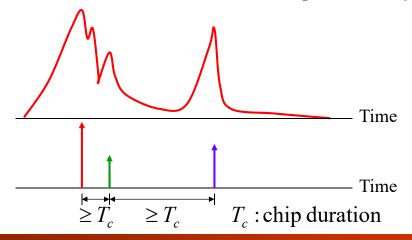
Frequency Diversity

- Frequency diversity:
 - Using multiple frequency channels
 - Each channel is separated by at least the coherence bandwidth of the channel
 - Frequency hopping spread spectrum (FHSS) systems provide frequency diversity through the fast frequency hopping
 - ⇒ Each symbol is transmitted on multiple hops (carriers)



Multipath Diversity

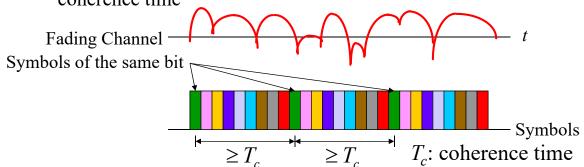
- Multipath diversity:
 - Resolving multipath components at different delays
 - Time resolution must be high enough \Rightarrow Wideband signals
 - Using direct sequence spread spectrum (DSSS) signaling along with a RAKE receiver can achieve multipath diversity



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Time Diversity

- Time diversity:
 - Using multiple time slots that are separated by at least the coherence time of the channel
 - Error correction coding with interleaving is an efficient method that provides time diversity
 - The coherence time of the channel depends on the **velocity** of the MS, and slow moving MSs have the channel with a large coherence time



Forward Error Correction Coding and Interleaving

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Forward Error Correction Coding

- Forward error correction (FEC) coding (or known as channel coding) is a technique used for controlling errors in data transmission
 - Over **noisy** and/or **unreliable (fading)** communication channels
- By introducing **redundancy** into an encoded message, the receiver can detect a limited number of errors that may occur anywhere in the message.
 - The **bandwidth efficiency** (throughput) is reduced, depending on the **code rate** $(0 \le r \le 1)$
 - For a large (small) code rate, the bandwidth efficiency is high (low), but the error-correction capability is low (high)
- FEC is also applied to **mass storage devices** to enable recovery of corrupted data and to improve the device reliability.

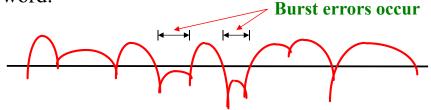
Forward Error Correction Coding

- There are two major categories of FEC codes:
 - Block codes: work on a fixed, predetermined number of bits or symbols (a fixed block size), such as Hamming codes, Reed-Solomon codes, BCH codes, LDPC codes, ...
 - The error-correction capability depends on both the **code rate** and the **block size**.
 - Convolutional codes: work on a bit or symbol stream of arbitrary length
 - The most often used decoding approach is soft decoding with the **Viterbi algorithm**.
 - The error-correction capability depends on both the **code rate** and the **constraint length** (the memory size in the encoder).
- FEC codes generally expects errors to be **uniformly** distributed.

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Interleaving

- **Interleaving** is frequently used in digital communications and storage systems to improve the performance of FEC codes.
- Many communication channels suffer **burst errors**, not random (independent) errors.
 - For an **AWGN channel**, the error pattern is random errors
 - For a **fading channel**, the error pattern is burst errors
- If the number of errors within a code word exceeds the errorcorrection capability, the decoder fails to recover the original code word.



Interleaving

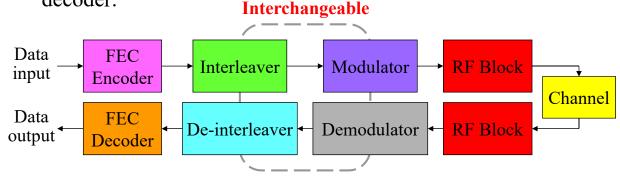
- Interleaving mitigates the impact of burst errors by shuffling contiguous coded symbols across time, frequency, or other domains
 - To make the burst errors become uniformly distributed errors.
- Considering a sequence of coded symbols with a block size of 8 symbols, each row of symbols corresponds to a codeword
 - Without interleaving, the transmission order is 1, 2, 3, 4, 5, 6, 7,
 8, 9, 10, ...
 Without interleaving

o, o, i o,								
With interleaving	1	2	3	4	5	6	7	8
	9	10	11	12	13	14	15	16
	17	18	19	20	21	22	23	24
	•	•	•	•	•	•	•	•
	449	450	451	452	453	454	455	456

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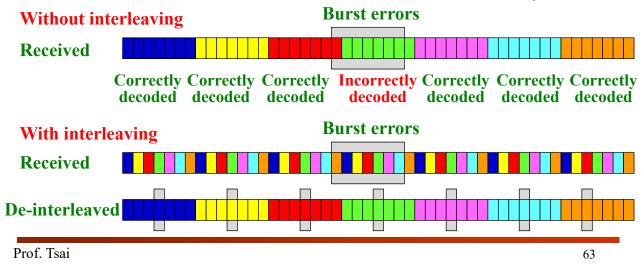
Interleaving

- With interleaving, the transmission order becomes 1, 9, 17, ..., 449, 2, 10, 18, ...
 - The adjacent symbols belong to different codewords.
 - The symbols of the same codeword are interleaved across time.
- At the transmitter, the interleaver is set after the FEC encoder.
- At the receiver, the de-interleaver is set in front of the FEC decoder.



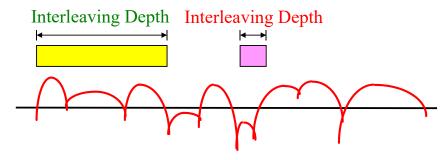
Interleaving

- When passing through the propagation channel, burst errors still occur in the received bit/symbol stream
 - Without interleaving, the receiver fails to decode some blocks
 - With interleaving, the burst errors becomes almost uniformly distributed errors and all blocks can be decoded correctly.



Interleaving

- The design of the **interleaving depth** relies on the **average envelope fade duration**.
- If the interleaving depth is smaller than the average envelope fade duration, all the symbols of a codeword may experience the same fade duration ⇒ burst errors occur in a codeword
- The interleaving depth must be larger than the average envelope fade duration.



Diversity Combining

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Diversity Combining

- Diversity combining refers to the method by which the signals from the diversity branches are combined
 - Predetection combining: diversity combining takes place <u>before</u> detection (takes place at RF)
 - Postdetection combining: diversity combining takes place <u>after</u> detection (takes place at baseband)
- For ideal coherent detection:
 - There is no difference in performance between predetection and postdetection combining

Diversity Combining

- Diversity combining methods:
 - Selective Combining (SC)
 - Maximal Ratio Combining (MRC)
 - Equal Gain Combining (EGC)
 - Switched Combining (SW)
- The received complex envelopes of the diversity branches:

$$\tilde{r}_k(t) = g_k \tilde{s}(t) + \tilde{n}_k(t), \quad k = 1, \dots, L$$

- $-g_k = \alpha_k e^{j\phi_k}$ is the complex fading gain associated with the k^{th} branch
- $-\widetilde{n}_k(t)$ is assumed to be independent from branch to branch
- The received signal vectors are

$$\tilde{\mathbf{r}}_k = g_k \tilde{\mathbf{s}} + \tilde{\mathbf{n}}_k, \quad k = 1, \dots, L$$

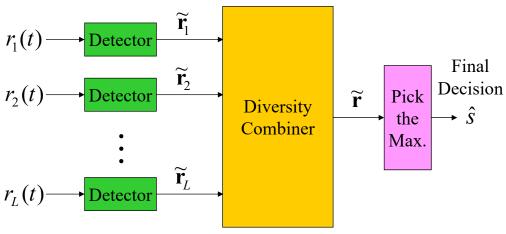
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Diversity Combining

- where $\tilde{\mathbf{r}}_k = \left[\tilde{r}_{k_1}, \tilde{r}_{k_2}, \cdots, \tilde{r}_{k_M}\right]$ is the received signal vector corresponding to **the** *M* **possible symbols**

$$\tilde{r}_{k_m} = g_k \tilde{s}_m + \tilde{n}_{k_m}, \quad m = 1, \dots, M$$

Postdetection combining



Fading Gains

- The fading gains of the various diversity branches typically have **some degree of correlation**, which depends on
 - The type of diversity being used
 - The propagation environment
- For analytical purposes, the diversity branches are generally assumed to be uncorrelated
 - Branch correlation reduces the achievable diversity gain
 - The uncorrelated branch assumption gives optimistic results

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Fade Distribution

- The fade distribution will affect the diversity gain
 - The relative advantage of diversity is greater for Rayleigh fading than Ricean fading
 - Ricean factor K increases \Rightarrow less difference between the instantaneous received SNR on the various diversity branches
 - The performance will always be better with Ricean than Rayleigh fading

Predetection Selective Combining (SC)

- The branch with the highest signal-to-noise ratio is selected at any instant
- With Rayleigh fading, the instantaneous received bit energy-tonoise ratio on *k*-th diversity branch has the exponential pdf

$$p_{\gamma_k}(x) = \frac{1}{\bar{\gamma}_c} e^{-x/\bar{\gamma}_c}$$

- $-\overline{\gamma}_c$ is the average received bit energy-to-noise ratio for each diversity branch
- With SC, the branch with the **largest** bit energy-to-noise ratio is always selected

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Predetection Selective Combining (SC)

• The effective instantaneous bit energy-to-noise ratio is

$$\gamma_s^S = \max\{\gamma_1, \gamma_2, \cdots, \gamma_L\}$$

- -L is the number of diversity branches
- If the diversity branches are independently faded, the cumulative distribution function (cdf) of γ_s^S is

$$F_{\gamma_k}(x) = \int_0^x \frac{1}{\bar{\gamma}_c} \exp(-\frac{y}{\bar{\gamma}_c}) \, dy = -\exp(-\frac{y}{\bar{\gamma}_c}) \Big|_0^x = 1 - e^{-x/\bar{\gamma}_c}$$

$$F_{\gamma_s^S}(x) = P_r \Big[\gamma_1 \le x, \, \gamma_2 \le x, \, \dots, \, \gamma_L \le x \Big] = \Big[1 - e^{-x/\bar{\gamma}_c} \Big]^L$$

• Differentiating the cdf gives the pdf

$$p_{\gamma_s^S}(x) = \frac{L}{\overline{\gamma}_c} \left[1 - e^{-x/\overline{\gamma}_c} \right]^{L-1} e^{-x/\overline{\gamma}_c}$$

Predetection Selective Combining (SC)

The average bit energy-to-noise ratio with SC is

$$\overline{\gamma}_{s}^{S} = \int_{0}^{\infty} x \, p_{\gamma_{s}^{S}}(x) \, dx$$

$$= \int_{0}^{\infty} \frac{L \, x}{\overline{\gamma}_{c}} \left[1 - e^{-x/\overline{\gamma}_{c}} \right]^{L-1} e^{-x/\overline{\gamma}_{c}} \, dx$$

$$= \overline{\gamma}_{c} \sum_{k=1}^{L} \frac{1}{k}$$

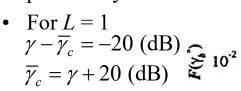
- The largest diversity gain is obtained by using 2-branch diversity:
 - $-L=1 \rightarrow 2 \Rightarrow 1.5/1 \Rightarrow 1.761 \text{ dB}$
 - $-L=2 \rightarrow 3 \Rightarrow 1.83/1.5 \Rightarrow 0.8715 \text{ dB}$

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Predetection Selective Combining (SC)

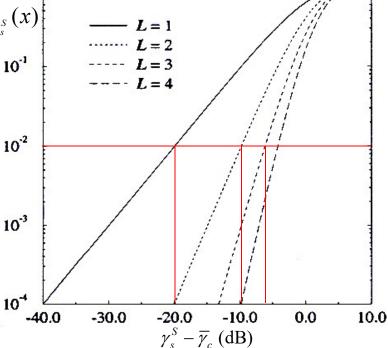
• Assume the desire SNR is γ and the $F_{\gamma_s^s}(x)$ acceptable outage probability is 0.01

10° 10-1



• For L=2 $\gamma - \overline{\gamma}_c = -10 \text{ (dB)}$ $\overline{\gamma}_c = \gamma + 10 \text{ (dB)}$

• For L = 3 $\gamma - \overline{\gamma}_c = -6 \text{ (dB)}$ $\overline{\gamma}_c = \gamma + 6 \text{ (dB)}$



Predetection Selective Combining (SC)

• The DPSK bit error probability with differential detection is:

$$P_b(\gamma_s) = \frac{1}{2}e^{-\gamma_s}$$

• With SC, the bit error probability for <u>DPSK is:</u>

Binomial expansion
$$P_{b} = \int_{0}^{\infty} P_{b}(x) p_{\gamma_{s}^{S}}(x) dx$$

$$= \int_{0}^{\infty} \frac{1}{2} \exp(-x) \frac{L}{\overline{\gamma_{c}}} [1 - \exp(-\frac{x}{\overline{\gamma_{c}}})]^{L-1} \exp(-\frac{x}{\overline{\gamma_{c}}}) dx$$

$$= \int_{0}^{\infty} \frac{L}{2\overline{\gamma_{c}}} \exp[-(1 + \frac{1}{\overline{\gamma_{c}}})x] [1 - \exp(-\frac{x}{\overline{\gamma_{c}}})]^{L-1} dx$$

$$= \frac{L}{2\overline{\gamma_{c}}} \sum_{n=0}^{L-1} C_{n}^{L-1} (-1)^{n} \int_{0}^{\infty} \exp[-(1 + \frac{1}{\overline{\gamma_{c}}} + \frac{n}{\overline{\gamma_{c}}})x] dx$$

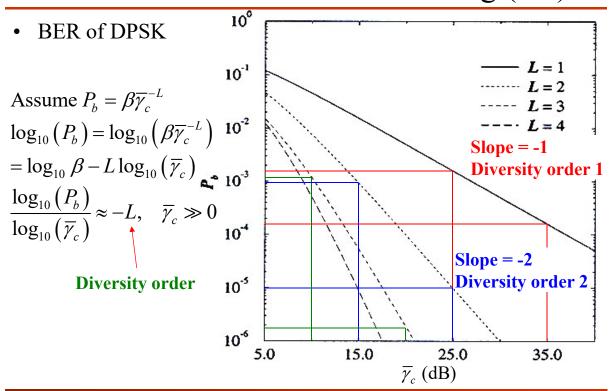
$$= \frac{L}{2\overline{\gamma_{c}}} \sum_{n=0}^{L-1} C_{n}^{L-1} (-1)^{n} \frac{\overline{\gamma_{c}}}{\overline{\gamma_{c}} + n + 1} = \frac{L}{2} \sum_{n=0}^{L-1} \frac{C_{n}^{L-1} (-1)^{n}}{\overline{\gamma_{c}} + n + 1}$$

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Predetection Selective Combining (SC)

- Diversity offers a very large improvement in the BER performance
- The bit error probability is now proportional to $\bar{\gamma}_c^{-L}$, not proportional to $\bar{\gamma}_c^{-1}$
- The largest diversity gain is achieved with 2-branch diversity

Predetection Selective Combining (SC)



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Maximal Ratio Combining (MRC)

- For Maximal Ratio Combining (MRC), the diversity branches must be **weighted** by their respective **complex fading gains** and then combined
- MRC results in a maximum likelihood (ML) receiver giving the <u>best possible performance</u> among the diversity combining techniques
- The ML receiver must have **complete knowledge** of the channel gain vector {**g**}
- Postdetection combining:
 - The weighting and combining is performed after integration
- Predetection combining:
 - The weighting and combining is performed before integration

• The received signal vector:

$$\tilde{\mathbf{r}} \triangleq (\tilde{r}_1, \tilde{r}_2, \dots, \tilde{r}_L); \quad \tilde{r}_k(t) \triangleq g_k \tilde{s}(t) + \tilde{n}_k(t)$$

• The ML receiver chooses the symbol \tilde{s}_m that maximizes the metric ML criterion

$$\mu_{1}(\tilde{s}_{m}) = -\sum_{k=1}^{L} |\tilde{r}_{k} - g_{k}\tilde{s}_{m}|^{2} = -\left(\sum_{k=1}^{L} |\tilde{r}_{k}|^{2} - 2\operatorname{Re}\left\{\left(\tilde{r}_{k}, g_{k}\tilde{s}_{m}\right)\right\} + |g_{k}\tilde{s}_{m}|^{2}\right)$$

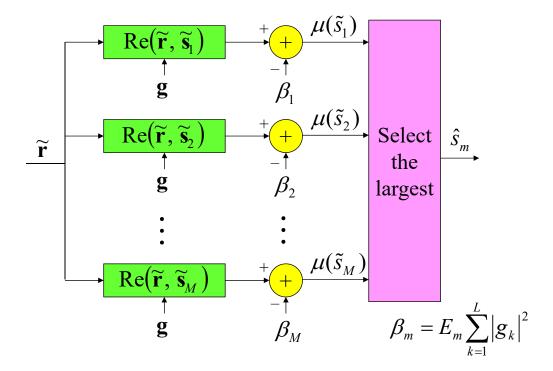
$$\mu_{2}(\tilde{s}_{m}) = \sum_{k=1}^{L} \operatorname{Re}\left\{\int_{0}^{T} \tilde{r}_{k}(t) \times g_{k}^{*}\tilde{s}_{m}^{*}(t) dt\right\} - \sum_{k=1}^{L} |g_{k}|^{2} E_{m}$$

$$= \sum_{k=1}^{L} \operatorname{Re}\left\{g_{k}^{*} \int_{0}^{T} \tilde{r}_{k}(t) \tilde{s}_{m}^{*}(t) dt\right\} - E_{m} \sum_{k=1}^{L} |g_{k}|^{2}$$
May be different for different S_{m}

The inner product:
$$(u, v) = \int u(t) \times v^*(t) dt$$
 $\|\tilde{s}_m\|^2 = 2E_m$

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Metric Computer for MRC



• If all symbols have equal energy, the metric becomes

$$\mu_3(\tilde{s}_m) = \sum_{k=1}^L \operatorname{Re} \left\{ g_k^* \int_0^T \tilde{r}_k(t) \tilde{s}_m^*(t) dt \right\}$$

• An alternative form of the ML receiver can also be obtained by rewriting the metric

$$\mu_4(\tilde{s}_m) = \int_0^T \operatorname{Re}\left\{\sum_{k=1}^L g_k^* \tilde{r}_k(t) \times \tilde{s}_m^*(t)\right\} dt - E_m \sum_{k=1}^L |g_k|^2$$

• Equivalently, the signal used for demodulation is

$$\tilde{r}(t) = \sum_{k=1}^{L} g_k^* \tilde{r}_k(t) = \sum_{k=1}^{L} g_k^* \tilde{g}_k \tilde{s}(t) + \sum_{k=1}^{L} g_k^* \tilde{n}_k(t)$$

$$g_k = \alpha_k e^{j\phi_k}; \quad g_k^* g_k = \alpha_k^2$$
Channel gain

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Maximal Ratio Combining (MRC)

• The channel-gain envelope of the composite signal is

$$\alpha_C = \sum_{k=1}^L \alpha_k^2$$

• Assuming that all branches have the same noise power, the sum of the branch **noise powers** is Weighted by

$$\sigma_{\widetilde{n},tot}^2 = N_0 \sum_{k=1}^L \alpha_k^2$$
 the channel gains

The symbol energy-to-noise ratio with MRC is

$$\gamma_s^{MR} = \frac{\alpha_C^2 E_{av}}{\sigma_{\tilde{n},tot}^2} = \sum_{k=1}^L \left(\alpha_k^2 \sum_{j=1}^L \alpha_j^2 E_{av}\right) / \left(N_0 \sum_{j=1}^L \alpha_j^2\right) = \sum_{k=1}^L \frac{\alpha_k^2 E_{av}}{N_0} = \sum_{k=1}^L \gamma_k$$
- where E_{av} is the average received symbol energy and

Average received symbol energy

$$\gamma_k = \alpha_k^2 E_{av} / N_0$$

- γ_s^{MR} is the sum of the bit energy-to-noise ratios of all diversity branches
- If all diversity branches provide the same average power and the branches are uncorrelated, then γ_s^{MR} has a chi-square distribution with 2L degrees of freedom

$$p_{\gamma_s^{MR}}(x) = \frac{1}{(L-1)!(\overline{\gamma}_c)^L} x^{L-1} e^{-x/\overline{\gamma}_c}$$

- where

$$\bar{\gamma}_c = E[\gamma_k], \quad k = 1, \dots, L$$

• The cdf of γ_s^{MR} :

$$F_{\gamma_s^{MR}}(x) = 1 - e^{-x/\overline{\gamma}_c} \sum_{k=0}^{L-1} \frac{1}{k!} \left(\frac{x}{\overline{\gamma}_c}\right)^k$$

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Maximal Ratio Combining (MRC)

• The average symbol energy-to-noise ratio with MRC is

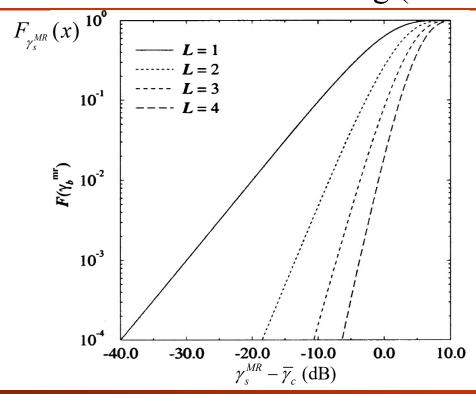
$$\overline{\gamma_s}^{MR} = E\left[\gamma_s^{MR}\right] = E\left[\sum_{k=1}^L \gamma_k\right] = \sum_{k=1}^L E\left[\gamma_k\right] = \sum_{k=1}^L \overline{\gamma}_k = \sum_{k=1}^L \overline{\gamma}_c = L\overline{\gamma}_c$$

- For SC, L = 2: $F_{\gamma_c^S}(x) = 10^{-4}$ at $\gamma_s^S \overline{\gamma}_c = -20$ dB
- For MRC, $L = 2 : F_{\gamma_s^{MR}}(x) = 10^{-4} \text{ at } \gamma_s^{MR} \overline{\gamma}_c = -18 \text{ dB}$
 - MRC is 2 dB more effective than SC
- The bit error probability of BPSK is

$$P_{b} = \int_{0}^{\infty} P_{b}(x) p_{\gamma_{s}^{MR}}(x) dx = \int_{0}^{\infty} Q(\sqrt{2x}) \frac{1}{(L-1)! (\overline{\gamma}_{c})^{L}} x^{L-1} e^{-x/\overline{\gamma}_{c}} dx$$

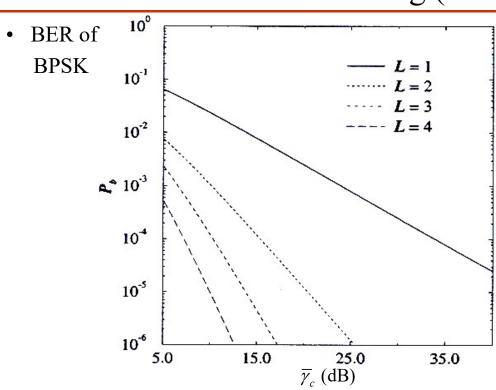
$$= \left(\frac{1-\mu}{2}\right)^{L} \sum_{k=0}^{L-1} C_{k}^{L-1+k} \left(\frac{1+\mu}{2}\right)^{k}$$

$$\mu = \sqrt{\frac{\overline{\gamma}_{c}}{1+\overline{\gamma}_{c}}}$$



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Maximal Ratio Combining (MRC)



Coherent Equal Gain Combining (EGC)

- The **coherent** Equal Gain Combining (EGC) is similar to MRC, but the diversity branches are **not weighted**
- EGC is useful for the modulation techniques having <u>equal</u> <u>energy symbols</u> (no channel information *g* is required), e.g. M-PSK
- For signals with unequal energy, the channel vector **g** is required and MRC (**best possible performance**) should be used
- The ML receiver chooses the symbol \tilde{s}_m that maximizes the metric

 Phase information

$$\mu(\tilde{s}_m) = \sum_{k=1}^{L} \operatorname{Re}\left\{\tilde{r}_k e^{-j\phi_k}, \tilde{s}_m\right\} = \sum_{k=1}^{L} \operatorname{Re}\left\{\int_0^T e^{-j\phi_k} \tilde{r}_k(t) \tilde{s}_m^*(t) dt\right\}$$

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Coherent Equal Gain Combining (EGC)

$$\mu(\tilde{s}_m) = \operatorname{Re}\left\{\sum_{k=1}^{L} e^{-j\phi_k} \tilde{r}_k, \tilde{s}_m\right\} = \int_0^T \operatorname{Re}\left\{\left(\sum_{k=1}^{L} e^{-j\phi_k} \tilde{r}_k(t)\right) \tilde{s}_m^*(t)\right\} dt$$

- For pre-detection combining, the signal after the diversity combiner is $\tilde{\mathbf{r}}(t) = \sum_{k=1}^{L} e^{-j\phi_k} \tilde{r}_k(t) = \sum_{k=1}^{L} e^{-j\phi_k} g_k \tilde{s}(t) + \sum_{k=1}^{L} e^{-j\phi_k} \tilde{n}_k(t)$
- The vector $\tilde{\mathbf{r}}$ is then applied to the metric computer for MRC with $\beta_m = 0, m = 1, \dots, L$
- The channel-gain envelope of the composite signal is

$$\alpha_E = \sum_{k=1}^{L} \alpha_k$$
 The symbol energy E_m is excluded

• Assuming that all branches have the same noise power, the sum of the branch **noise powers** is LN_0

Coherent Equal Gain Combining (EGC)

• The symbol energy-to-noise ratio with EGC is

$$\gamma_s^{EG} = \frac{\alpha_E^2 E_{av}}{LN_0}$$

- where E_{av} is the average symbol energy
- The cdf and pdf of γ_s^{EG} do not exist in closed form for L > 2
- For L=2 and $\overline{\gamma}_1=\overline{\gamma}_2=\overline{\gamma}_c$, the cdf and pdf are

$$F_{\gamma_s^{EG}}(x) = 1 - e^{-2x/\overline{\gamma}_c} - \sqrt{\pi} \frac{x}{\overline{\gamma}_c} e^{-x/\overline{\gamma}_c} \left(1 - 2Q \left(\sqrt{2 \frac{x}{\overline{\gamma}_c}} \right) \right)$$

$$p_{\gamma_s^{EG}}(x) = \frac{1}{\overline{\gamma}_c} e^{-2x/\overline{\gamma}_c} + \sqrt{\pi} e^{-x/\overline{\gamma}_c} \left(\frac{1}{2\sqrt{x\overline{\gamma}_c}} - \frac{1}{\overline{\gamma}_c} \sqrt{\frac{x}{\overline{\gamma}_c}} \right) \times \left(1 + 2Q \left(\sqrt{2 \frac{x}{\overline{\gamma}_c}} \right) \right)$$

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Coherent Equal Gain Combining (EGC)

• The average symbol energy-to-noise ratio with EGC is

$$\overline{\gamma_s^{EG}} = \frac{E_{av}}{LN_0} E \left[\alpha_E^2\right] = \frac{E_{av}}{LN_0} E \left[\left(\sum_{k=1}^L \alpha_k\right)^2\right] = \frac{E_{av}}{LN_0} \sum_{k=1}^L \sum_{\ell=1}^L E\left[\alpha_k \alpha_\ell\right]$$

• For an uncorrelated Rayleigh fading channel:

$$E\left[\alpha_{k}^{2}\right] = 2\sigma^{2} \text{ and } E\left[\alpha_{k}\alpha_{\ell}\right] = E\left[\alpha_{k}\right]E\left[\alpha_{\ell}\right] = \left(\sqrt{\pi/2}\sigma\right)^{2} = \pi\sigma^{2}/2$$

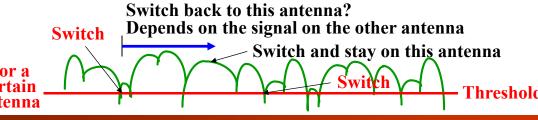
$$\overline{\gamma}_{s}^{EG} = \frac{E_{av}}{LN_{0}}\left(2L\sigma^{2} + L(L-1)\frac{\pi\sigma^{2}}{2}\right)$$

$$= \frac{2\sigma^{2}E_{av}}{N_{0}}\left(1 + (L-1)\frac{\pi}{4}\right) = \overline{\gamma}_{c}\left(1 + (L-1)\frac{\pi}{4}\right) \leq L\overline{\gamma}_{c}$$

• The bit error probability of coherent BPSK (2-branch):

$$P_{b} = \int_{0}^{\infty} P_{b}(x) p_{\gamma_{s}^{EG}}(x) dx = \frac{1}{2} \left(1 - \sqrt{1 - \mu^{2}} \right), \quad \mu = \frac{1}{1 + \overline{\gamma}_{c}}$$

- A switched combiner scans through the diversity branches until it finds one that has a signal-to-noise ratio exceeding a specified threshold
- This diversity branch is used until the signal-to-noise ratio again **drops below the threshold**
- Two-branch Switch and Stay Combining (SSC):
 - The receiver switches to, and stays with, the alternate branch when the bit energy-to-noise ratio drops below the threshold
 - It does this regardless of whether or not the bit energy-to-noise ratio of the alternate branch is **above or below** the threshold



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Predetection Switched Combining (SW)

- Assume that the symbol energy-to-noise ratios associated with the two branches are γ_1 and γ_2 and the switching threshold is T
- The pdf of the received bit energy-to-noise ratio is

$$p_{\gamma_k}(x) = \frac{1}{\overline{\gamma}_c} e^{-x/\overline{\gamma}_c}$$

Define

$$q = \Pr[\gamma_i < T] = 1 - e^{-T/\overline{\gamma}_c}$$
$$p = \Pr[\gamma_i \le S] = 1 - e^{-S/\overline{\gamma}_c}$$

• The symbol energy-to-noise ratio at the output of the switched combiner is γ_s^{SW}

$$\Pr\left[\gamma_{s}^{SW} \leq S\right] = \Pr\left[\gamma_{s}^{SW} \leq S \middle| \gamma_{s}^{SW} = \gamma_{1}\right] \cup \Pr\left[\gamma_{s}^{SW} \leq S \middle| \gamma_{s}^{SW} = \gamma_{2}\right]$$

• γ_1 is statistically identical to $\gamma_2 \Rightarrow$ Assuming that branch 1 is currently in use, we have the cdf of γ_s^{SW} expressed as

$$\Pr \Big[\gamma_s^{SW} \leq S \Big] = \begin{cases} \Pr \Big[\big\{ \gamma_1 < T \big\} \bigcap \big\{ \gamma_2 \leq S \big\} \big], & S < T \\ \Pr \Big[\big\{ T \leq \gamma_1 \leq S \big\} \bigcup \big\{ \gamma_1 < T \bigcap \gamma_2 \leq S \big\} \big], & S \geq T \end{cases}$$

• Since γ_1 and γ_2 are independent, we have

$$\Pr \Big[\big\{ \gamma_1 < T \big\} \cap \big\{ \gamma_2 \le S \big\} \Big] = qp$$

$$\Pr \Big[\big\{ T \le \gamma_1 \le S \big\} \cup \big\{ \gamma_1 < T \cap \gamma_2 \le S \big\} \Big] = \Big(p - q \Big) + qp$$

$$\Pr \Big[\gamma_s^{SW} \le S \Big] = \begin{cases} qp, & S < T \\ p - q + qp, & S \ge T \end{cases}$$

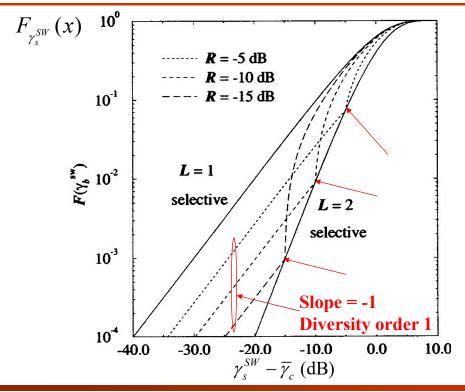
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Predetection Switched Combining (SW)

• The normalized threshold is defined as

$$R = 10\log_{10}(T/\bar{\gamma}_c)(dB)$$

- SSC always performs worse than SC
 - Except at the switching threshold, where the performance is the same as SC
- The threshold level T should be chosen as γ_{th}
 - The minimum acceptable instantaneous symbol energy-to-noise ratio
- The optimum threshold, $T = R\overline{\gamma}_c$, depends on $\overline{\gamma}_c$
 - Since $\bar{\gamma}_c$ varies due to path loss and shadowing, the normalized threshold R must be adaptive



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Predetection Switched Combining (SW)

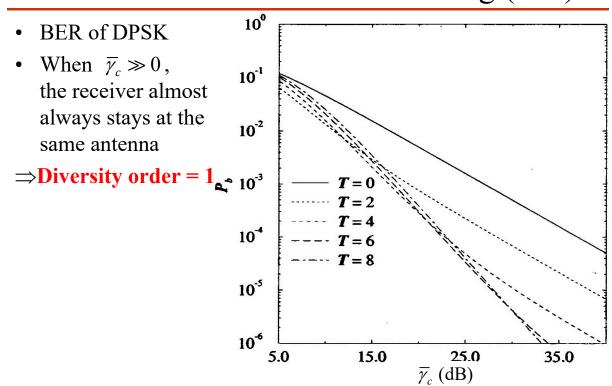
• The pdf of γ_s^{SW} is

$$p_{\gamma_s^{SW}}(x) = \begin{cases} q \frac{1}{\overline{\gamma}_c} e^{-x/\overline{\gamma}_c}, & x < T \\ (1+q) \frac{1}{\overline{\gamma}_c} e^{-x/\overline{\gamma}_c}, & x \ge T \end{cases}$$

• If binary DPSK is used, the probability of error is

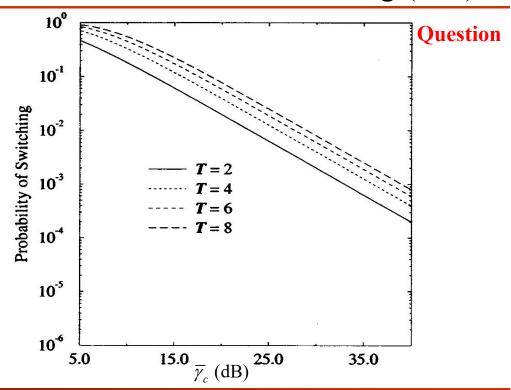
$$P_b = \int_0^\infty P_b(x) p_{\gamma_s^{SW}}(x) dx$$
 Diversity order = 1
$$= \frac{1}{2(1+\overline{\gamma}_c)} \left(\underline{q + (1-q)e^{-T}} \right)$$
 Almost a constant for $\overline{\gamma}_c \gg 0$

- The performance with T = 0 (no switching): no diversity at all
- The performance changes little for T > 6



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Predetection Switched Combining (SW)



Transmit Diversity

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Transmit Diversity (TD)

- Transmit diversity uses **multiple transmit antennas** to provide the receiver with multiple uncorrelated replicas of the same signal
 - The complexity of having multiple antennas may be shared among many receivers
 - Generally used in **forward link** (base-to-mobile)
- The receiver with only **one** antenna can still benefit from a diversity gain

Space-Time Transmit Diversity

- Assume that two transmit antennas and one receiver antenna are used
- In any given baud period, two data symbols are transmitted simultaneously from the two transmit antennas
 - The symbols transmitted from Antenna 1 and 2 are $\widetilde{\mathbf{s}}_{(1)}$ and $\widetilde{\mathbf{s}}_{(2)}$
 - During the next baud period, the symbols transmitted from Antenna 1 and 2 are $-\widetilde{\mathbf{s}}_{(2)}^*$ and $\widetilde{\mathbf{s}}_{(1)}^*$
- The channel gains for the two antennas are $g_1(t)$ and $g_2(t)$
- Assume that the channel stays constant over two baud intervals

$$g_k(t) = g_k(t+T) = g_k = \alpha_k e^{j\phi_k}$$

− where *T* is the baud period

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Space-Time Transmit Diversity

The received complex signal vectors are:

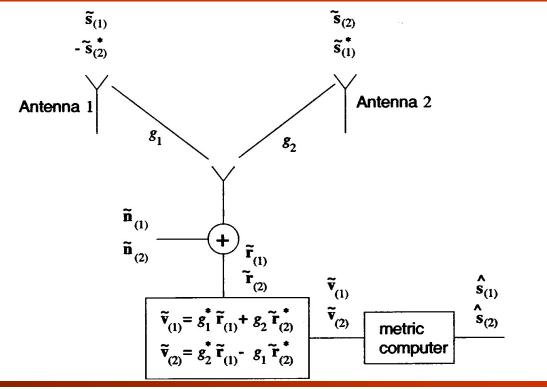
$$\tilde{\mathbf{r}}_{(1)} = g_1 \tilde{\mathbf{s}}_{(1)} + g_2 \tilde{\mathbf{s}}_{(2)} + \tilde{\mathbf{n}}_{(1)}
\tilde{\mathbf{r}}_{(2)} = -g_1 \tilde{\mathbf{s}}_{(2)}^* + g_2 \tilde{\mathbf{s}}_{(1)}^* + \tilde{\mathbf{n}}_{(2)}$$

- where $\widetilde{\mathbf{r}}_{(1)}$ and $\widetilde{\mathbf{r}}_{(2)}$ represent the received vectors at t and t+T
- The diversity combiner constructs the following two signal vectors: $\tilde{\mathbf{v}}_{m} = g_{*}^{*} \tilde{\mathbf{r}}_{m} + g_{*} \tilde{\mathbf{r}}_{m}^{*}$

$$\tilde{\mathbf{v}}_{(1)} = g_1^* \tilde{\mathbf{r}}_{(1)} + g_2 \tilde{\mathbf{r}}_{(2)}^*
\tilde{\mathbf{v}}_{(2)} = g_2^* \tilde{\mathbf{r}}_{(1)} - g_1 \tilde{\mathbf{r}}_{(2)}^*
\tilde{\mathbf{v}}_{(1)} = (\alpha_1^2 + \alpha_2^2) \tilde{\mathbf{s}}_{(1)} + g_1^* \tilde{\mathbf{n}}_{(1)} + g_2 \tilde{\mathbf{n}}_{(2)}^*
\tilde{\mathbf{v}}_{(2)} = (\alpha_1^2 + \alpha_2^2) \tilde{\mathbf{s}}_{(2)} - g_1 \tilde{\mathbf{n}}_{(2)}^* + g_2^* \tilde{\mathbf{n}}_{(1)}
g_1^* g_1 = (\alpha_1 e^{-j\phi_1}) (\alpha_1 e^{j\phi_1}) = \alpha_1^2$$

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Space-Time Transmit Diversity (2×1 Diversity)



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Space-Time Transmit Diversity

• Applying the metric computer for MRC, we can make the decisions by maximizing the metric

$$\mu(\tilde{\mathbf{s}}_{(1),m}) = \text{Re}(\tilde{\mathbf{v}}_{(1)}, \tilde{\mathbf{s}}_{(1),m}) - E_m(|g_1|^2 + |g_2|^2)$$

$$\mu(\tilde{\mathbf{s}}_{(2),m}) = \text{Re}(\tilde{\mathbf{v}}_{(2)}, \tilde{\mathbf{s}}_{(2),m}) - E_m(|g_1|^2 + |g_2|^2)$$

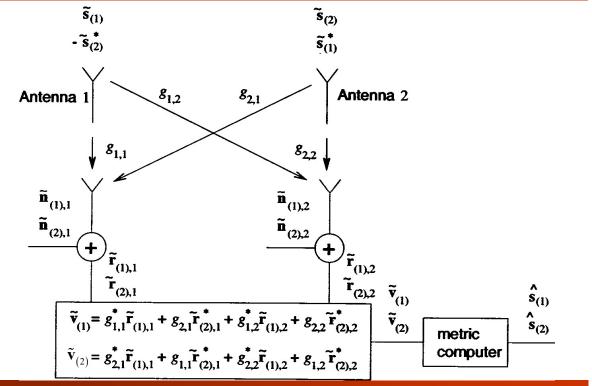
• Compared with the output of the MRC with L = 2

$$\tilde{\mathbf{r}} = g_1^* \tilde{\mathbf{r}}_1 + g_2^* \tilde{\mathbf{r}}_2$$

$$= (\alpha_1^2 + \alpha_2^2) \tilde{\mathbf{s}}_m + g_1^* \tilde{\mathbf{n}}_1 + g_2^* \tilde{\mathbf{n}}_2$$

• The combined signals are the same, except for the **phase** rotations of the noise vectors

Space-Time Transmit Diversity (2×2 Diversity)



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Signal Space Diversity

Signal Space Diversity (SSD)

- The SSD technology was proposed to exploit **diversity gain** for the transmission through fading channels.
- At the **transmitter**, the input information bits are modulated to form a set of complex symbols, such as M-PSK or QAM.
- Subsequently, the modulated signals are phase rotated by a predetermined angle and followed by an ideal component interleaving process.
 - Interleave the I-part and Q-part components of multiple symbols to form a set of new symbols, in each of which the two components come from two original symbols.
- Then the new (**interleaved** and **combined**) symbols are transmitted through the propagation channels.

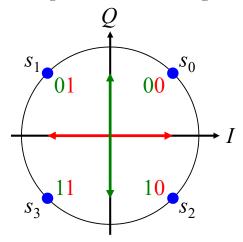
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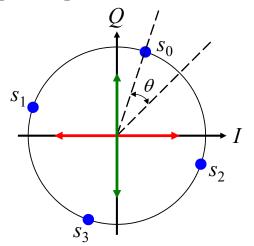
Signal Space Diversity (SSD)

- At the receiver, the signals received on all channels are component de-interleaved to reconstruct the symbols corresponding to the original symbols.
- Multiple original symbols are involved in using the common channels
 - All related symbols are jointly detected based on the maximum likelihood (ML) criterion.
- If L original symbols are involved in SSD, an diversity order L can be achieved.
- No additional power or bandwidth resources are required.

Signal Space Diversity (SSD)

- Without phase rotation, I-part/Q-part depends only on one bit
- With phase rotation, I-part/Q-part depends on the two bits



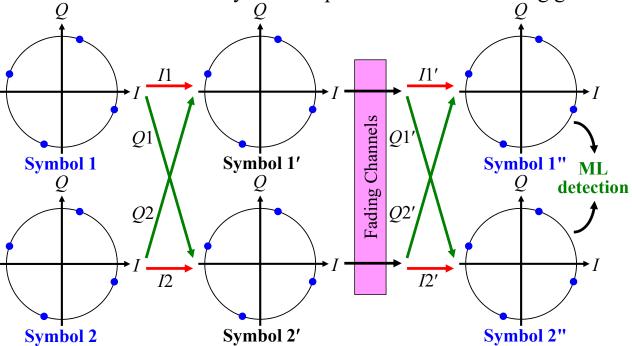


- \longleftrightarrow I-part: representing the bit 0 or 1
- \longrightarrow Q-part: representing the bit 0 or 1
- → I-part: representing the two bits
- → Q-part: representing the two bits

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Signal Space Diversity (SSD)

• The two transmitted symbols experience different fading gains



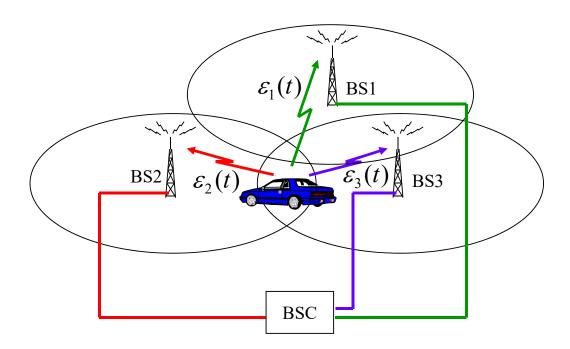
Macroscopic Diversity

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Macroscopic Diversity

- Microscopic diversity applies some diversity techniques to mitigate the effects of envelope fading (fast fading)
- Macroscopic diversity is a diversity technique that is used to combat the effects of shadow fading
 - Up-link signals from a mobile station are received by two or more geographically separated base stations
 - The received signals are diversity combined
 - The diversity advantage is obtained if the signals experience some degree of uncorrelated shadowing
- Assume that there are *M* diversity branches (BSs)
- The average received bit energy-to-noise ratio per diversity branch is $\bar{\gamma}_c$ (in dB)

Macroscopic Diversity



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Macroscopic Diversity

• We have
$$p(\overline{\gamma}_{c(dB)}) = \frac{1}{\sqrt{2\pi}\sigma_{\overline{\gamma}_c}} \exp\left\{-\left(\overline{\gamma}_{c(dB)} - \mu_{\overline{\gamma}_c}\right)^2 / 2\sigma_{\overline{\gamma}_c}^2\right\}$$

- where $\mu_{\bar{\gamma}_c} = E[\bar{\gamma}_{c(dB)}]$
- The most effective and simplest method for macroscopic diversity is selective combining
 - The BS that provides the largest average received bit energy-tonoise ratio is selected as the serving BS
- Let $\overline{\gamma}_{th(dB)}$ be a specified threshold level, and $\overline{\gamma}_{ck(dB)}$ be the average received bit energy-to-noise ratio from BS k

$$\begin{split} \Pr(\overline{\gamma}_{ck(\mathrm{dB})} \leq \overline{\gamma}_{th(\mathrm{dB})}) &= \int_{-\infty}^{\overline{\gamma}_{th(\mathrm{dB})}} \frac{1}{\sqrt{2\pi}\sigma_{\overline{\gamma}_{c}}} \exp\left\{-\left(\overline{\gamma}_{ck(\mathrm{dB})} - \mu_{\overline{\gamma}_{ck}}\right)^{2} \middle/ 2\sigma_{\overline{\gamma}_{c}}^{2}\right\} d\overline{\gamma}_{ck} \\ &= Q\left(\left(\mu_{\overline{\gamma}_{ck}} - \overline{\gamma}_{th(\mathrm{dB})}\right) \middle/ \sigma_{\overline{\gamma}_{c}}\right) \end{split}$$

Macroscopic Diversity

- Let $\mathbf{d} = (d_0, d_1, \dots, d_{M-1})$ be the set of distances between the MS and the *M* different BSs
- The outage probability of thermal noise:

$$O(\mathbf{d}) = \prod_{k=0}^{M-1} \Pr(\bar{\gamma}_{ck(\mathrm{dB})} \leq \bar{\gamma}_{th(\mathrm{dB})}) = \prod_{k=0}^{M-1} Q \left(\frac{\mu_{\bar{\gamma}_{ck}} - \bar{\gamma}_{th(\mathrm{dB})}}{\sigma_{\bar{\gamma}_c}} \right)$$

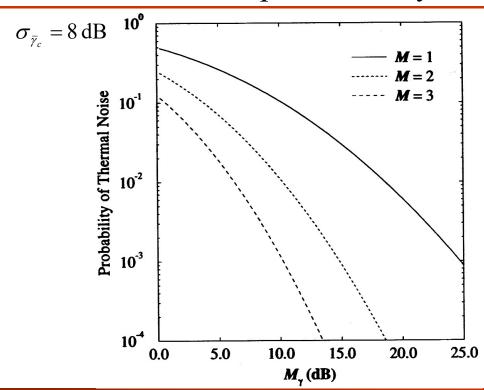
- Assume that $\mu_{\bar{\gamma}_{ck}}$ are all equal to $\mu_{\bar{\gamma}_c}$, i.e., $d_0 = d_1 = \dots = d_{M-1}$
 - The outage probability becomes

$$O = \left[Q \left(\frac{\mu_{\overline{\gamma}_c} - \overline{\gamma}_{th(dB)}}{\sigma_{\overline{\gamma}_c}} \right) \right]^M$$

• The thermal noise margin: $M_{\gamma} = \mu_{\bar{\gamma}_c} - \bar{\gamma}_{th(\text{dB})}$

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Macroscopic Diversity

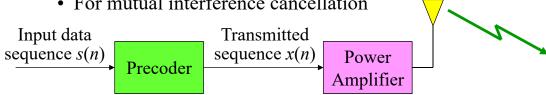


Precoding

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Precoding

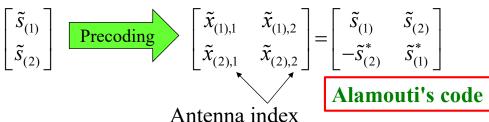
- **Precoding** is a modification of a sequence of signals (symbols) before transmission to meet a specific goal/requirement.
- There are various purposes for using precoding in wireless transmissions
 - To decrease the envelope variations of the transmitted signals
 - To mitigate the inter-symbol interference induced by channels
 - To achieve radio resource mapping for different signals
 - For diversity reception
 - For multiple access
 - For mutual interference cancellation



Precoding – Transmit Diversity

- Space-time transmit diversity is an example of using precoding
- The input data signals are transformed to new modified signals for transmission
 - To achieve diversity reception at the receiver
- For 2×1 transmit diversity, the transmitted signals are

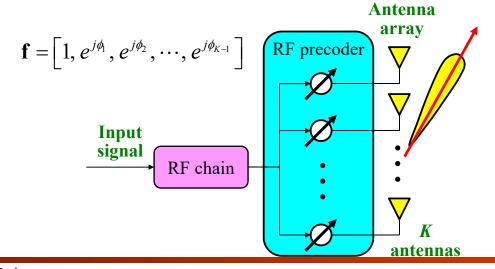
Original data signals Transmitted signals



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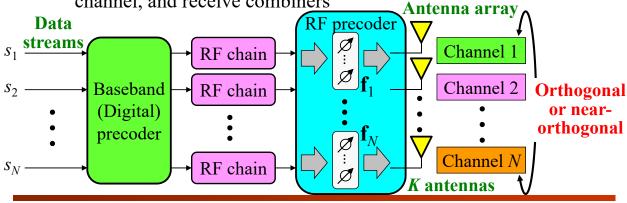
Precoding – Beamforming

- The same signal is emitted from each of the transmit antennas with appropriately chosen **weighting** (**phase** and/or **gain**)
 - Such that the signal power is **maximized** in a desired direction
 - For RF precoding, a constant amplitude constraint is required



Precoding – Spatial Multiplexing

- Multiple data streams are emitted from the transmit antennas with different and appropriately chosen weightings
 - Such that the transmitter can simultaneously transmit multiple data streams to the receiver/receivers using the same resources
 - The goal is to **maximize** the total system throughput
 - **Multiple spatial channels** are formed through the precoders, channel, and receive combiners



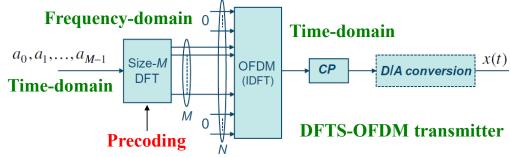
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Precoding – DFT-Spread OFDM

- Flexible bandwidth assignment: For OFDM-based uplink transmission, dynamically allocating different number of subcarriers to different terminals depending on their instantaneous channel conditions
- To achieve the goal of **low-PAPR** (peak-to-average power ratio) at a user equipment, "**single-carrier**" transmission is used
 - The **DFT-spread OFDM** (DFTS-OFDM) transmission scheme
 - Small variations in the instantaneous signal power
 - Possibility for low-complexity high-quality equalization in the frequency domain
 - Possibility for FDMA with flexible bandwidth assignment

Precoding – DFT-Spread OFDM

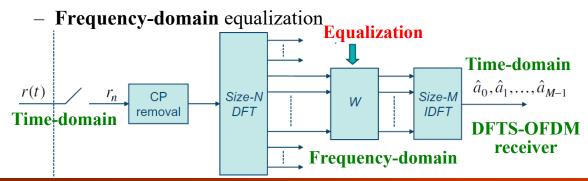
- The DFTS-OFDM transmission can be regarded as normal OFDM transmission with a DFT-based precoding
 - A block of M modulation symbols (BPSK, QPSK, 16QAM, ...)
 is first applied to a size-M DFT
 - The output is then applied to consecutive inputs (subcarriers) of an OFDM modulator, depending on the bandwidth assignment
 - The OFDM modulator is a size-N inverse DFT (IDFT) with N >
 M and the unused inputs of the IDFT are set to zero



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Precoding – DFT-Spread OFDM

- The demodulation of a DFTS-OFDM signal is similar to the demodulation of an OFDM signal
 - Using size-N DFT processing with removal of the frequency samples not corresponding to the desired signal
 - Using size-M inverse DFT processing to obtain data estimate
- An equalizer is needed to compensate for the radio-channel frequency selectivity



Multi-Antenna Techniques

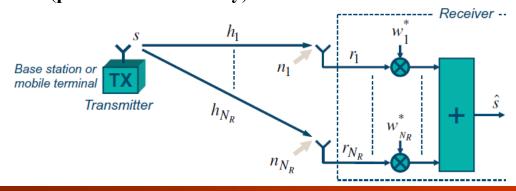
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Antenna Distance

- An important characteristic of multi-antenna configuration is the **distance** between the different antenna elements
 - The mutual correlation between the radio-channel fading experienced by the signals at the different antennas
 - A **large** inter-antenna distance \Rightarrow a **low** mutual fading correlation
 - Almost independent fast fading
 - A small inter-antenna distance \Rightarrow a high mutual fading correlation
 - Very similar fast fading
- Whether **high or low correlation** is desirable depends on what is to be achieved with the multi-antenna configuration
 - Diversity, beamforming, or spatial multiplexing

Benefits of Multi-Antenna: Diversity

- Using multiple antennas at the transmitter and/or the receiver to provide additional diversity against channel fading
 - Low mutual correlation between different antennas, implying
 - The need for a sufficiently large inter-antenna distance (spatial diversity) or
 - The use of different antenna polarization directions (polarization diversity)

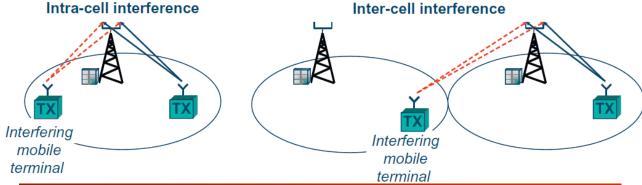


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Benefits of Multi-Antenna: Beamforming

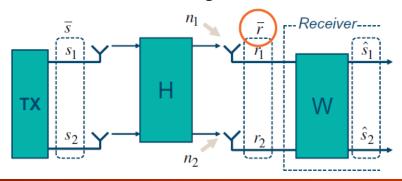
- Using multiple antennas at the transmitter **and/or** the receiver to "**shape**" the overall **antenna beam pattern**
 - Very high mutual correlation or low mutual correlation between different antennas
 - **Beamforming** to **maximize** the overall antenna gain in the direction of the target receiver/transmitter

• Beamforming to suppress specific dominant interference cell interference Inter-cell interference



Benefits of Multi-Antenna: Spatial Multiplexing

- **Simultaneously** using multiple antennas at the transmitter and the receiver to support **multiple parallel channels** over the radio interface with the same radio resources
 - Spatial multiplexing to achieve very high bandwidth efficiency
 - Also referred to as MIMO (Multi-Input Multi-Output) antenna processing
- The case of 2×2 antenna configuration



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Benefits of Multi-Antenna: Spatial Multiplexing

• The received signals can be expressed as:

$$\mathbf{r} = \begin{bmatrix} r_1 \\ r_2 \end{bmatrix} = \begin{bmatrix} h_{1,1} & h_{1,2} \\ h_{2,1} & h_{2,2} \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} + \begin{bmatrix} n_1 \\ n_2 \end{bmatrix} = \mathbf{H}\mathbf{s} + \mathbf{n}$$

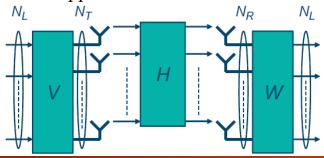
• Assuming no noise and that the channel matrix \mathbf{H} is **invertible**, both signals s_1 and s_2 can be **perfectly** recovered at the receiver

$$\hat{\mathbf{s}} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} = \mathbf{W}\mathbf{r} = \mathbf{H}^{-1}\mathbf{r} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} + \mathbf{\underline{H}^{-1}n} \quad \text{Noise}$$

- The properties of **H** will determine the resultant noise level
 - The closer the channel matrix is to being a singular matrix, the larger the increase in the noise level

Benefits of Multi-Antenna: Spatial Multiplexing

- Consider a multiple-antenna configuration consisting of N_T transmit antennas and N_R receive antennas.
- The number of parallel signals that can be spatially multiplexed is **upper limited** by $N_L = \min\{N_T, N_R\}$
 - No more than N_T different signals can be transmitted from N_T transmit antennas
 - With N_R receive antennas, a maximum of N_R 1 interfering signals can be suppressed



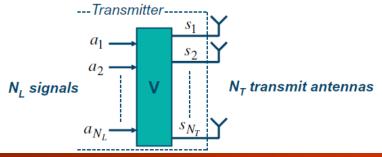
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Benefits of Multi-Antenna: Spatial Multiplexing

- However, in many cases, the number of spatially multiplexed signals will be **less than** N_L
- In the case of **very bad** channel conditions (low signal-to-noise ratio, SNR) there is **no gain** of spatial multiplexing
 - The channel capacity is a linear function of the SNR
 - The multiple antennas should be used for **beamforming** to improve the SNR, rather than for spatial multiplexing
- The spatial-multiplexing order should be determined based on the properties of the size $N_R \times N_T$ channel matrix
 - Any excess antennas should be used to provide beamforming
 - Such combined beamforming and spatial multiplexing can be achieved by means of precoder-based spatial multiplexing

Precoder-Based Spatial Multiplexing

- Linear precoding for spatial multiplexing is achieved by using a size $N_T \times N_L$ precoding matrix at the transmitter side
- In the general case with $N_L \le N_T$, N_L signals are spatially multiplexed and transmitted using N_T transmit antennas
 - If $N_L = N_T$, the precoding can be used to "**orthogonalize**" the parallel transmissions, allowing for improved signal isolation
 - If $N_L < N_T$, the precoding provides the mapping including the combination of **spatially multiplexing** and **beamforming**



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Precoder-Based Spatial Multiplexing

- In the LTE system, the codebook-based precoding allows for a maximum of **four antenna ports** (output signals) and, as a consequence, a maximum of **four layers** (input signals)
- Precoder matrices for two antenna ports and one and two layers

One layer
$$\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \mathbf{W}x$$

$$\frac{1}{\sqrt{2}} \begin{bmatrix} +1 \\ +1 \end{bmatrix} \qquad \frac{1}{\sqrt{2}} \begin{bmatrix} +1 \\ -1 \end{bmatrix} \qquad \frac{1}{\sqrt{2}} \begin{bmatrix} +1 \\ +j \end{bmatrix} \qquad \frac{1}{\sqrt{2}} \begin{bmatrix} +1 \\ -j \end{bmatrix}$$

$$- \text{Two layers} \qquad \mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \mathbf{W}\mathbf{x} = \mathbf{W} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\frac{1}{\sqrt{2}} \begin{bmatrix} +1 & 0 \\ 0 & +1 \end{bmatrix} \qquad \frac{1}{2} \begin{bmatrix} +1 & +1 \\ +1 & -1 \end{bmatrix} \qquad \frac{1}{2} \begin{bmatrix} +1 & +1 \\ +j & -j \end{bmatrix}$$

Equalization

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Adaptive Equalizer

- An adaptive equalizer is an adjustable filters at the receiver
 - Used to mitigate the combined effect of ISI and noise
- Two broad categories of equalizers:
 - Symbol-by-symbol equalizers
 - Include a decision device to make symbol-by-symbol decisions on the received symbol sequence
 - Sequence estimators
 - Make decisions on a sequence of received symbols
- Sequence estimators are generally more complex than symbolby-symbol equalizers, but can potentially offer better performance

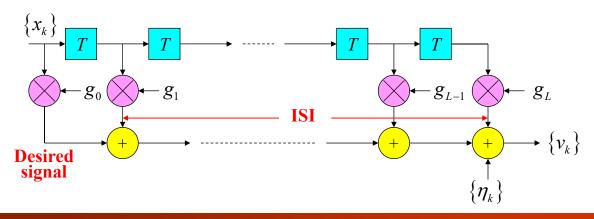
Modeling of ISI Channels

• The multipath channel can be represented by a **discrete-time transversal filter** with coefficients:

$$\mathbf{g} = \left[g_0, g_1, \dots, g_{L-1}, g_L\right]^T$$

• If the channel gain is time-variant, the coefficients become

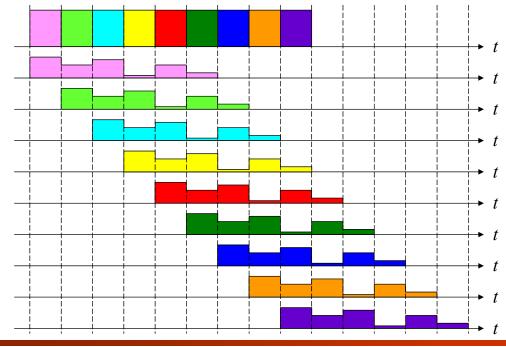
$$\mathbf{g}(k) = [g_0(k), g_1(k), \dots, g_{L-1}(k), g_L(k)]^T$$



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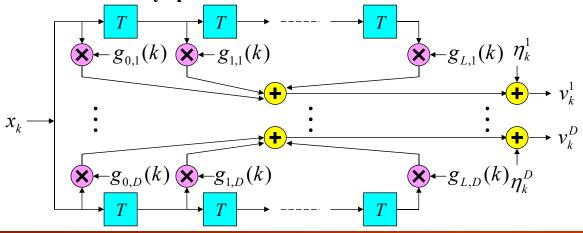
Modeling of ISI Channels

• Each input symbol triggers a channel impulse response in time



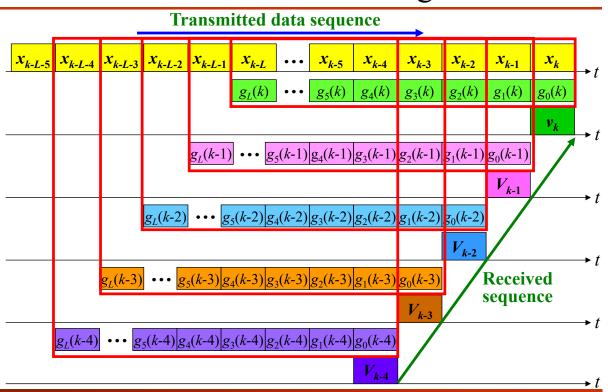
MLSE and the Viterbi Algorithm

- MLSE: maximum likelihood sequence estimation
- The overall channel can be modeled by a collection of D transversal filters that are T-spaced and have L+1 taps
 - -D is the number of **diversity branches**
 - -L is the delay spread



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MLSE and the Viterbi Algorithm



MLSE and the Viterbi Algorithm

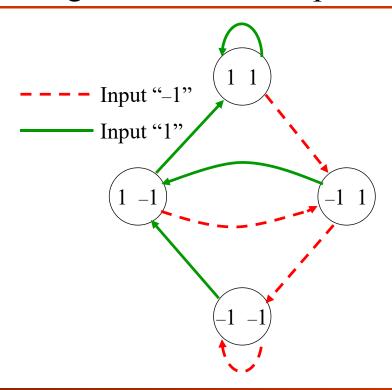
- The channel has a finite number of states
 - If the size of the signal constellation is 2^n (*M*-ary modulation)
 - There are a total of $N_s = 2^{nL}$ states (L shift registers)
- The state at epoch k is:

$$S_k = (x_{k-1}, x_{k-2}, \dots, x_{k-L})$$

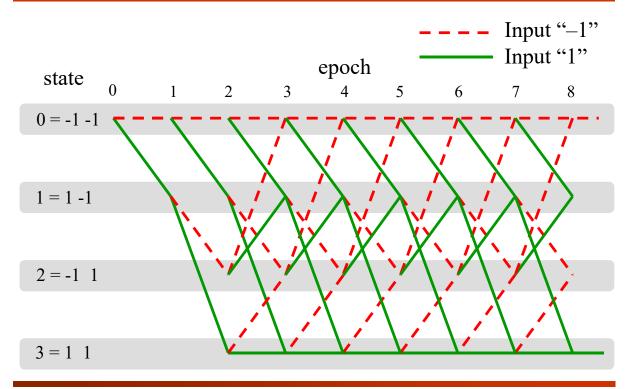
- For example: the binary sequence $\mathbf{x}, x_n \in \{-1, +1\}$, is transmitted over a three-tap static ISI channel with channel vector $\mathbf{g} = (1, 1, 1)$
- There are four states (3-tap \Rightarrow 2 shift registers \Rightarrow 2² = 4 states)
- The system can be described by the state diagram
- The system state diagram can be used to construct the **trellis** diagram
 - The initial zero state is assumed to be $S_0^{(0)} = (-1, -1)$

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State Diagram for Three-Tap ISI Channel



Trellis Diagram for Three-Tap ISI Channel



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MLSE and the Viterbi Algorithm

• Let the vector of signals received on all diversity branches at epoch *n* be

$$\mathbf{V}_{n} = (v_{n,1}, v_{n,2}, ..., v_{n,D})$$

• After receiving the sequence $\{V_n\}_{n=1}^k$, the ML receiver decides in favor of the sequence $\{x_n\}_{n=1}^k$ that maximizes the **likelihood** function or **log-likelihood function**

$$p(\mathbf{V}_k, ..., \mathbf{V}_1 | x_k, ..., x_1), \log p(\mathbf{V}_k, ..., \mathbf{V}_1 | x_k, ..., x_1)$$

• Possible input data sequences:

$$\mathbf{x}_{1} = (x_{k,1}, x_{k-1,1}, ..., x_{2,1}, x_{1,1}) \Rightarrow p_{1} = p(\mathbf{V}_{k}, ..., \mathbf{V}_{1} \mid \mathbf{x}_{1})$$

$$\mathbf{x}_{2} = (x_{k,2}, x_{k-1,2}, ..., x_{2,2}, x_{1,2}) \Rightarrow p_{2} = p(\mathbf{V}_{k}, ..., \mathbf{V}_{1} \mid \mathbf{x}_{2})$$

$$\vdots$$

$$\mathbf{x}_{M^{k}} = (x_{k,M^{k}}, x_{k-1,M^{k}}, ..., x_{2,M^{k}}, x_{1,M^{k}}) \Rightarrow p_{M^{k}} = p(\mathbf{V}_{k}, ..., \mathbf{V}_{1} \mid \mathbf{x}_{M^{k}})$$

MLSE and the Viterbi Algorithm

• Find the maximum value of p_i (or $\log p_i$)

$$\hat{\mathbf{x}} = \arg \max_{\mathbf{x}_j, 1 \le j \le M^k} \left\{ p(\mathbf{V}_k, \dots, \mathbf{V}_1 \mid \mathbf{x}_j) \right\}$$

• Since V_n depends only on the L most recent transmitted symbols, the log-likelihood function at epoch k is denoted as

$$\log p(\mathbf{V}_{k}, \dots, \mathbf{V}_{1} \mid x_{k}, \dots, x_{1})$$

$$= \log p(\mathbf{V}_{k} \mid x_{k}, \dots, x_{k-L}) + \underline{\log p(\mathbf{V}_{k-1}, \dots, \mathbf{V}_{1} \mid x_{k-1}, \dots, x_{1})}$$

$$- \text{ where } x_{k-L} = 0 \text{ for } k - L \le 0$$

- Assume that the second term of the right-hand side has been calculated previously at epoch k-1
 - Only the first term (branch metric) has to be computed for each incoming signal vector \mathbf{V}_k at epoch k

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MLSE and the Viterbi Algorithm

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New item to be calculated

Has been calculated

MLSE and the Viterbi Algorithm

• The conditional pdf of the received signal vector:

$$p(\mathbf{V}_{k} \mid x_{k}, \dots, x_{k-L}) = \frac{1}{(\pi N_{0})^{D}} \exp \left\{ -\frac{1}{N_{0}} \sum_{d=1}^{D} \left| v_{k,d} - \sum_{i=0}^{L} g_{i,d} x_{k-i} \right|^{2} \right\}$$

• For $\log p(\mathbf{V}_k | x_k, ..., x_{k-L})$, the **branch metric** is

$$\mu_{k} = -\sum_{d=1}^{D} \left| v_{k,d} - \hat{v}_{k,d} \left(x_{k}, \dots, x_{k-i} \right) \right|^{2} = -\sum_{d=1}^{D} \left| v_{k,d} - \sum_{i=0}^{L} g_{i,d} x_{k-i} \right|^{2}$$

- The <u>Viterbi algorithm</u> can be used to implement the ML receiver by searching through the N_s -state trellis for most likely transmitted sequence \mathbf{x}
- This search process is called maximum likelihood sequence estimation (MLSE)

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MLSE and the Viterbi Algorithm

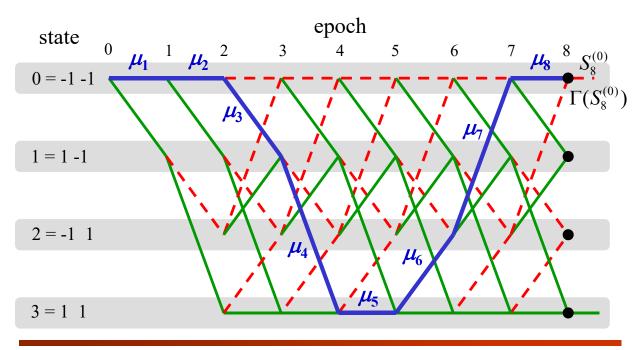
- At epoch k, assume that the algorithm has stored N_s surviving sequences $\check{\mathbf{x}}(S_k^{(i)})$ (paths through the trellis)
 - with associated **path metrics** $\Gamma(S_k^{(i)})$ (distances from the received sequence)

$$\Gamma(S_k^{(i)}) = \sum_{\{k\}} \mu_k$$

- $\{\mu_k\}$ is the set of **branch metrics** along the surviving path $\breve{\mathbf{x}}(S_k^{(i)})$

Trellis Diagram for Three-Tap ISI Channel

: a surviving path



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MLSE and the Viterbi Algorithm

- After the vector V_k has been received, the Viterbi algorithm executes the following steps for each state $S_{k+1}^{(j)}$, $j = 0, ..., N_s 1$
 - Compute the set of path metrics for all possible paths through the trellis that terminate in state $S_{k+1}^{(j)}$

$$\Gamma(S_k^{(i)} \to S_{k+1}^{(j)}) = \Gamma(S_k^{(i)}) + \mu(S_k^{(i)} \to S_{k+1}^{(j)})$$

- Find $\Gamma(S_{k+1}^{(j)}) = \max \Gamma(S_k^{(i)} \to S_{k+1}^{(j)})$, where the maximization is over all possible paths through the trellis that terminate in $S_{k+1}^{(j)}$
- Store $\Gamma(S_{k+1}^{(j)})$ and its associated **surviving sequence** $\breve{\mathbf{x}}(S_{k+1}^{(j)})$, and **drop all other paths**Symbols in
- $\mu(S_k^{(i)} \to S_{k+1}^{(j)})$ is the branch metric associated with the the surviving sequence transition $S_k^{(i)} \to S_{k+1}^{(j)}$ and is computed as

$$\mu(S_k^{(i)} \to S_{k+1}^{(j)}) = -\sum_{d=1}^{D} \left| v_{k,d} - g_{0,d} x_k (S_k^{(i)} \to S_{k+1}^{(j)}) - \sum_{m=1}^{L} g_{m,d} \underline{x_{k-m}} (S_k^{(i)}) \right|^2$$

MLSE and the Viterbi Algorithm

- $x_k(S_k^{(i)} \to S_{k+1}^{(j)})$ is a symbol that is uniquely determined by the transition $S_k^{(i)} \to S_{k+1}^{(j)}$, and the L most recent symbols $\{x_{k-m}(S_k^{(i)})\}_{m=1}^{L}$ are uniquely specified by the previous states $S_k^{(i)}$
- When two paths merged into a state, why only the path with a larger path metric is stored (as the surviving path)?
- Consider the two sequences s1="1001100" and s2="0101000" merged at State 0, with the path metric M1 > M2
 - Assume that the subsequent sequence with the maximum path metric is "...101101"
 - Then, the metrics "1001100101101" > "0101000101101" > "0101000XXXXXXX" \Rightarrow So, the sequence s2 can be discarded
- After all states have been processed, the time index k is incremented and the whole algorithm repeats

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MLSE and the Viterbi Algorithm

- Theoretically, the ML receiver waits until the entire sequence $\{V_n\}_{n=1}^{\infty}$ has been received before making a decision
- In practice, a decision about x_{k-Q} is usually made when V_k is received and processed
 - If Q > 5L, the performance degradation caused by the resulting path metric truncation is negligible
- The decision delay shall be Q > 5L
 - Consider a sequence currently with the maximum path metric



Cumulative Path Metrics

- Suppose that the data sequence $\mathbf{x} = (-1, 1, 1, -1, 1, 1, -1, -1, ...)$ is transmitted
- The noiseless received sequence is $\mathbf{v} = (v_0, v_1, v_2, v_3, v_4, \dots)$

$$v_n = g_0 x_n + g_1 x_{n-1} + g_2 x_{n-2}$$

= $x_n + x_{n-1} + x_{n-2}$ $g = (1, 1, 1)$
 $S_0^{(0)} = (-1, -1)$

• Then the noiseless received sequence is

$$\mathbf{v} = (-3, -1, 1, 1, 1, 1, 1, -1, \cdots)$$

- where the previous two symbols are assumed to be (-1, -1)
- Assume that the noisy received sequence is

$$\mathbf{v} = (-3.2, -1.1, 0.9, 0.1, 1.2, 1.5, 0.7, -1.3, \cdots)$$

• The path metrics are equal to the square Euclidean distance between surviving sequence $\bar{\mathbf{x}}(S_k^{(i)})$ and received sequence \mathbf{v}

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Cumulative Path Metrics

$$\Gamma(S_0^{(0)}) = 0; \Gamma(S_0^{(1)}) = 0; \Gamma(S_0^{(2)}) = 0; \Gamma(S_0^{(3)}) = 0;$$

$$S_k^{(0)} \to S_{k+1}^{(0)} \Rightarrow v_{k+1} = -3; S_k^{(0)} \to S_{k+1}^{(1)} \Rightarrow v_{k+1} = -1;$$

$$S_k^{(1)} \to S_{k+1}^{(2)} \Rightarrow v_{k+1} = -1; S_k^{(1)} \to S_{k+1}^{(3)} \Rightarrow v_{k+1} = +1;$$

$$S_k^{(2)} \to S_{k+1}^{(0)} \Rightarrow v_{k+1} = -1; S_k^{(2)} \to S_{k+1}^{(1)} \Rightarrow v_{k+1} = +1;$$

$$S_k^{(3)} \to S_{k+1}^{(2)} \Rightarrow v_{k+1} = +1; S_k^{(3)} \to S_{k+1}^{(3)} \Rightarrow v_{k+1} = +3;$$

• Epoch 1

$$\mu(S_0^{(0)} \to S_1^{(0)}) = -|-3.2 - (-3)|^2 = -0.04; \implies \Gamma(S_0^{(0)} \to S_1^{(0)}) = -0.04$$

 $\mu(S_0^{(0)} \to S_1^{(1)}) = -|-3.2 - (-1)|^2 = -4.84; \implies \Gamma(S_0^{(0)} \to S_1^{(1)}) = -4.84$

• Epoch 2

$$\mu(S_1^{(0)} \to S_2^{(0)}) = -|-1.1 - (-3)|^2 = -3.61; \Gamma(S_1^{(0)} \to S_2^{(0)}) = -0.04 - 3.61 = -3.65$$

$$\mu(S_1^{(0)} \to S_2^{(1)}) = -|-1.1 - (-1)|^2 = -0.01; \Gamma(S_1^{(0)} \to S_2^{(1)}) = -0.04 - 0.01 = -0.05$$

$$\mu(S_1^{(1)} \to S_2^{(2)}) = -|-1.1 - (-1)|^2 = -0.01; \Gamma(S_1^{(1)} \to S_2^{(2)}) = -4.84 - 0.01 = -4.85$$

$$\mu(S_1^{(1)} \to S_2^{(3)}) = -|-1.1 - (+1)|^2 = -4.41; \Gamma(S_1^{(1)} \to S_2^{(3)}) = -4.84 - 4.41 = -9.25$$

Cumulative Path Metrics

• Epoch 3

$$\mu(S_2^{(0)} \to S_3^{(0)}) = -|0.9 - (-3)|^2 = -15.21; \Gamma(S_2^{(0)} \to S_3^{(0)}) = -3.65 - 15.21 = -18.86 \otimes \mu(S_2^{(2)} \to S_3^{(0)}) = -|0.9 - (-1)|^2 = -3.61; \Gamma(S_2^{(2)} \to S_3^{(0)}) = -4.85 - 3.61 = -8.46$$

$$\mu(S_2^{(0)} \to S_3^{(1)}) = -|0.9 - (-1)|^2 = -3.61; \Gamma(S_2^{(0)} \to S_3^{(1)}) = -3.65 - 3.61 = -7.26 \otimes \mu(S_2^{(2)} \to S_3^{(1)}) = -|0.9 - (+1)|^2 = -0.01; \Gamma(S_2^{(2)} \to S_3^{(1)}) = -4.85 - 0.01 = -4.86$$

$$\mu(S_2^{(1)} \to S_3^{(2)}) = -|0.9 - (-1)|^2 = -3.61; \Gamma(S_2^{(1)} \to S_3^{(2)}) = -0.05 - 3.61 = -3.66$$

$$\mu(S_2^{(3)} \to S_3^{(2)}) = -|0.9 - (+1)|^2 = -0.01; \Gamma(S_2^{(3)} \to S_3^{(2)}) = -9.25 - 0.01 = -9.26 \otimes \mu(S_2^{(1)} \to S_3^{(3)}) = -|0.9 - (+1)|^2 = -0.01; \Gamma(S_2^{(1)} \to S_3^{(2)}) = -0.05 - 0.01 = -0.06$$

$$\mu(S_2^{(3)} \to S_3^{(3)}) = -|0.9 - (+3)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -0.9 - (+3)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 = -4.41; \Gamma(S_2^{(3)} \to S_3^{(3)}) = -9.25 - 4.41 = -13.66 \otimes \mu(S_2^{(3)} \to S_3^{(3)}) = -1.09 - (-1)|^2 \to -1.09 + (-1)|^2 \to -$$

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Cumulative Path Metrics

• Epoch 4

$$\mu(S_3^{(0)} \to S_4^{(0)}) = -|0.1 - (-3)|^2 = -9.61; \Gamma(S_3^{(0)} \to S_4^{(0)}) = -8.46 - 9.61 = -18.07 \otimes \mu(S_3^{(2)} \to S_4^{(0)}) = -|0.1 - (-1)|^2 = -1.21; \Gamma(S_3^{(2)} \to S_4^{(0)}) = -3.66 - 1.21 = -4.87$$

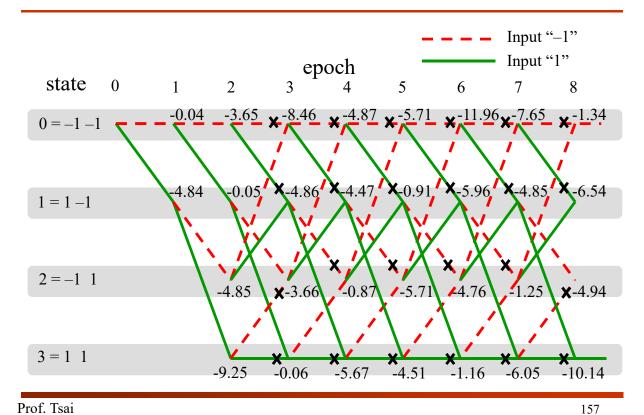
$$\mu(S_3^{(0)} \to S_4^{(1)}) = -|0.1 - (-1)|^2 = -1.21; \Gamma(S_3^{(0)} \to S_4^{(1)}) = -8.46 - 1.21 = -9.67 \otimes \mu(S_3^{(2)} \to S_4^{(1)}) = -|0.1 - (+1)|^2 = -0.81; \Gamma(S_3^{(2)} \to S_4^{(1)}) = -3.66 - 0.81 = -4.47$$

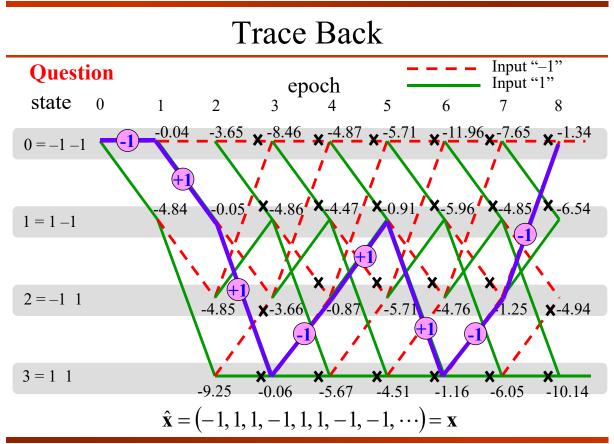
$$\mu(S_3^{(1)} \to S_4^{(2)}) = -|0.1 - (-1)|^2 = -1.21; \Gamma(S_3^{(1)} \to S_4^{(2)}) = -4.86 - 1.21 = -6.07 \otimes \mu(S_3^{(3)} \to S_4^{(2)}) = -|0.1 - (+1)|^2 = -0.81; \Gamma(S_3^{(3)} \to S_4^{(2)}) = -0.06 - 0.81 = -0.87$$

$$\mu(S_3^{(1)} \to S_4^{(3)}) = -|0.1 - (+1)|^2 = -0.81; \Gamma(S_3^{(1)} \to S_4^{(2)}) = -4.86 - 0.81 = -5.67$$

$$\mu(S_3^{(3)} \to S_4^{(3)}) = -|0.1 - (+3)|^2 = -8.41; \Gamma(S_3^{(3)} \to S_4^{(3)}) = -0.06 - 8.41 = -8.47 \otimes 1.00$$

Cumulative Path Metrics





Adaptive MLSE Receiver

- The Viterbi algorithm requires knowledge of the channel vectors \mathbf{g}_d (channel gain of each diversity branches) to compute the branch metrics
 - An adaptive channel estimator is needed
 - A transversal digital filter with LMS algorithm is used
- With the LMS (Least-Mean-Square) algorithm, the tap coefficients are updated by $\hat{g}_{i,d}(k+1) = \hat{g}_{i,d}(k) + \alpha \varepsilon_{\underline{k-Q},d} \hat{x}_{\underline{k-i-Q}}^*$, $i = 0, \dots, L; d = 1, \dots, D$
 - $-\alpha$ is the adaptation step size
 - $\varepsilon_{k-O,d}$ is the error associated with branch d, defined as

$$\varepsilon_{k-Q,d} = v_{k-Q,d} - \sum_{i=0}^{L} \hat{g}_{i,d}(k) \hat{x}_{k-i-Q}$$

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Adaptive MLSE Receiver

- A major problem with this channel estimator is:
 - It lags behind the true channel vector by the **decision delay** Q (in the Viterbi algorithm)
- Actually, the received signal at epoch k Q is

$$v_{k-Q,d} = \sum_{i=0}^{L} g_{i,d}(k-Q)x_{k-i-Q} + \eta_{k-Q,d}$$

- The signal $v_{k-Q,d}$ is influenced by $g_{i,d}(k-Q)$, not $g_{i,d}(k)$
- The error:

$$\varepsilon_{k-Q,d} = \sum_{i=0}^{L} (g_{i,d}(k-Q) - \hat{g}_{i,d}(k)) x_{k-i-Q} + \eta_{k-Q,d}$$

- which depends on the difference of $g_{i,d}(k-Q)$ (at **epoch** k-Q) and the estimated $\hat{g}_{i,d}(k)$ (at **epoch** k)

Adaptive MLSE Receiver

- The channel time variations over the decision delay Q will cause the term $\{g_{i,d}(k-Q) \hat{g}_{i,d}(k)\}_{k=1}^{L}$ to be non-zero, and degrade the tracking performance
- The decision delay Q could be reduced but this will also reduce the reliability of the decisions \hat{x}_{k-i-Q} that are used to update the channel estimates
- The overall performance improvement obtained by reducing *Q* is often **minimal**

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Adaptive MLSE Receiver

- The **Per-survivor processing** approach:
 - Each state has its own channel estimator that tracks the channel
 - Multiple channel estimators are required
 - The tap coefficients are updated according to $\hat{g}_{i,d}(k+1) = \hat{g}_{i,d}(k) + \alpha \varepsilon_{k,d} \ \breve{x}_{k-i}^*, \quad i = 0, \dots, L; \ d = 1, \dots, D$
 - where \mathbf{x} is the surviving sequence associated with **each state**
 - Each state uses zero-delay symbols ⇒ good channel tracking performance for the correct data sequence

Adaptive MLSE Receiver

- For the LMS algorithm:
 - Initialize the algorithm by setting

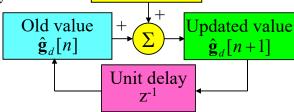
$$\hat{\mathbf{g}}_{d}[1] = [\hat{g}_{0,d}, \hat{g}_{1,d}, \cdots, \hat{g}_{L,d}] = \mathbf{0}$$

- For n = 1, 2, ..., it computes

$$\hat{\mathbf{v}}_{d}[n] = \mathbf{x}^{T}[n] \,\hat{\mathbf{g}}_{d}[n], \quad \mathbf{\varepsilon}_{d}[n] = \mathbf{v}_{d}[n] - \hat{\mathbf{v}}_{d}[n]$$
$$\hat{\mathbf{g}}_{d}[n+1] = \hat{\mathbf{g}}_{d}[n] + \alpha \mathbf{\varepsilon}_{d}[n] \,\mathbf{x}[n]$$

- Continue the iterative computation until the equalizer reaches a "steady state"

 Correction
- LMS algorithm is a feedback system
- \Rightarrow It is possible to **diverge**



 $\alpha \mathbf{\epsilon}[n] \mathbf{x}[n]$

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Training Sequence

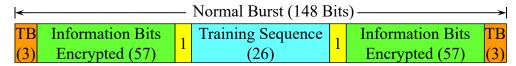
- The role of training sequence is to train the adaptive equalizer in **training mode**
- The training sequence is a **fixed-pattern** (**known**) sequence, for the adaptive equalizer to estimate the channel vector **g**
- For **TDMA** systems, the training sequence should be transmitted **in each slot**
- Long training sequences:
 - Good estimation of the channel
 - Wasting the spectrum resource
- **Short** training sequences:
 - Bad estimation of the channel \Rightarrow degrade the performance

Training Sequence

- The position of the training sequence could be in:
 - The **head** of a frame
 - The **middle** of a frame

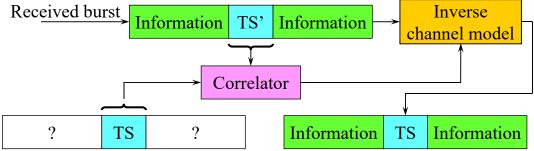
Question

- The **tail** of a frame
- For GSM systems, the training sequence is located in the **middle** of a burst
 - The length of the training sequence is 26 bits
 - There are a total of 8 training sequence codes (TSC)
- GSM systems use the maximum likelihood sequence estimation (MLSE) with Viterbi algorithm
- It reflects the maximum delay of 15 μs (4 bits)



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Training Sequence



Link Adaptation

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Link Adaptation

- Link adaptation is a technique used to improve the transmission efficiency in wireless communications
 - Adapt the modulation, coding and other parameters to the transmission conditions (e.g., the propagation loss, the interference, the sensitivity of a receiver, the available transmit power margin, etc.).
- Link adaptation systems generally require some **channel state information (CSI)** at the transmitter.
 - For TDD systems, the transmitter can approximate the CSI of the Tx-to-Rx link as the CSI estimate of the Rx-to-Tx link
 - Under the reciprocal theorem and slow varying assumption
 - For FDD systems, the receiver needs to feed back the CSI estimate to the transmitter
 - It may only feed back the decoding results (Success/Failure) or the received signal strength (RSS)

Link Adaptation

- The most popular link adaptation scheme adapts only modulation and coding
- According to the **channel quality**, the transmitter jointly choose the modulation type and code rate for transmission
 - If the transmission condition is **favorable**, the transmitter uses high-order modulation and high-rate coding
 - To improve the throughput
 - If the transmission condition is unfavorable, the transmitter may downgrade the modulation order and/or the code rate
 - To ensure successful reception
- If high-order modulation and high-rate coding are used in an **unfavorable** transmission condition
 - The packet error rate will be very high and the throughput is still degraded

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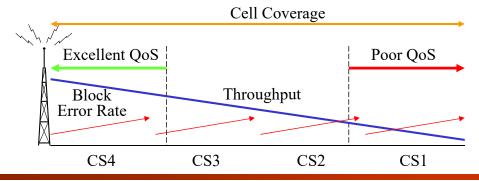
Link Adaptation (GPRS)

- For example, GPRS (General Packet Radio Service) in GSM systems uses a rate adaptation algorithm that adapts the coding scheme (CS) according to the quality of the radio channel
 - Modulation: GMSK (1 bit/symbol);
 - Code rate: 1.0, 3/4, 2/3, 1/2;
- The code rate adaptation is achieved by bit puncturing
 - The **same encoder** (code rate 1/2) is used for different code rates

Scheme	Code Rate	Punctured bits	Data rate (kbps)
CS-1	1/2	0	9.05
CS-2	2/3	132	13.4
CS-3	3/4	220	15.6
CS-4	1	-	21.4

Link Adaptation (GPRS)

- When a receiver is close to the BS (with an **excellent channel quality**), high-rate transmission (e.g., CS4) is used
 - If low-rate transmission is used, throughput will be degraded
- When a receiver is at the cell border (with a **poor channel quality**), low-rate transmission (e.g., CS1) is used
 - If high-rate transmission (e.g., CS4) is used, throughput is also degraded because of a very high packet error probability



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Link Adaptation (EDGE)

- For example, EDGE (Enhanced Data rates for GSM Evolution) uses a rate adaptation algorithm that adapts the modulation and coding scheme (MCS) according to the quality of the radio channel
 - Modulation: GMSK (1 bit/symbol); 8PSK (3 bits/symbol);
 - Code rate: 1.0, 0.8, 0.66, 0.53 (GMSK); 1.0, 0.92, 0.76, 0.49, 0.37 (8PSK);

Scheme	Code rate	Modulation	Data rate (kb/s)	Scheme	Code rate	Modulation	Data rate (kb/s)
MCS-4	1.0	- GMSK	17.6	MCS-9	1.0	8PSK	59.2
MCS-3	0.80		14.8/13.6	MCS-8	0.92		54.4
MCS-2	0.66		11.2	MCS-7	0.76		44.8
MCS-1	0.53		8.8	MCS-6	0.49		29.6/27.2
					0.37		22.4

Link Adaptation (HSDPA)

- HSDPA (High-Speed Downlink Packet Access) in 3G CDMA systems uses a rate adaptation algorithm that adapts the modulation and coding scheme, known as the AMC (Adaptive Modulation and Coding) scheme, according to the quality of the radio channel
 - Modulation: QPSK (2 bits/symbol); 8-PSK (3 bits/symbol); 16-QAM (4 bits/symbol); 64-QAM (6 bits/symbol);
 - Code rate: 3/4, 1/2, 1/4 (QPSK); 3/4 (8-PSK); 3/4, 1/2 (16-QAM);
 3/4 (64-QAM);
- Traditionally, CDMA systems have used **fast power control** as the preferred method for link adaptation.
 - Control the transmission power, not control the transmission rate
- The AMC scheme has offered an alternative link adaptation method that promises to raise the **overall system capacity**.

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Link Adaptation (HSDPA)

- With AMC, the power of the transmitted signal is held constant over a frame interval
 - The modulation and coding format is changed to match the current received signal quality or channel conditions.

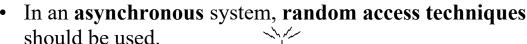
Scheme	Modulation	Code Rate	Data rate (Mbps)
MCS1	QPSK	1/4	1.2
MCS2	QPSK	1/2	2.4
MCS3	QPSK	3/4	3.6
MCS4	8-PSK	3/4	5.4
MCS5	16-QAM	1/2	4.8
MCS6	16-QAM	3/4	7.2
MCS7	64-QAM	3/4	10.8

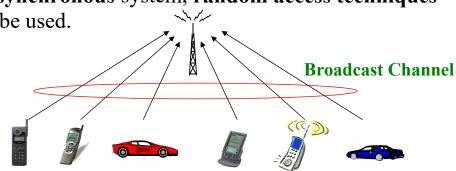
Random Access Techniques

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Random Access

- In a broadcast channel (such as a wireless channel), one of the key issue is to determine who gets the right of using the channel when there is **competition** for it.
- In a synchronous system (with a controller), the controlledaccess (multiple access) techniques, such as TDMA or CDMA, can be applied to prevent/reduce signal collision or mutual interference.





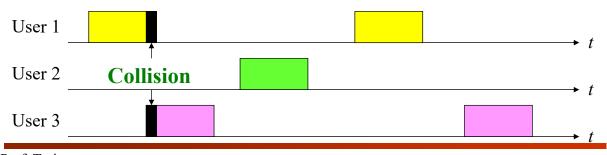
Random Access Schemes

- Random access techniques are completely decentralized
- Each user determines its transmission depends solely on a **local** random mechanism and the **local observation**
- It is possible that two or more users decide to transmit their signals in an **overlapping** time interval
 - It is called as a collision
- Several random access schemes have been proposed for wireless communications
 - Pure ALOHA
 - Slotted ALOHA
 - CSMA/CD
 - MACA

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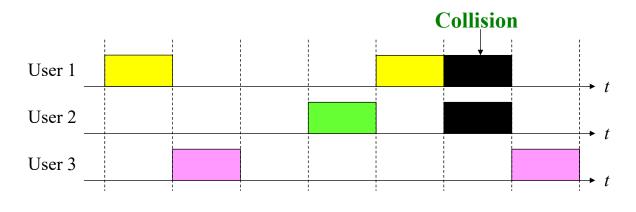
Pure ALOHA

- Pure ALOHA is the **simplest** random access scheme
- Each user transmits the data packet **immediately** whenever it has a packet for transmission
- There is a feedback mechanism to acknowledge the **success/failure** of the transmitted packet
 - Such as the positive acknowledgement with timeout
- When a collision occurs, the user will **re-transmit** the packet after a **random backoff time**



Slotted ALOHA

- In slotted ALOHA, the time scale is slotted into **multiple time slots** with the duration equal to the packet length
- Users can transmit data packets at the **beginning** of each time slot only
- The other processes are the same as the pure ALOHA scheme



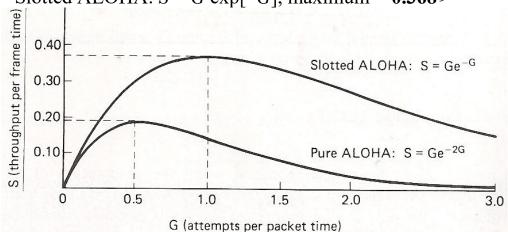
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ALOHA – Average Throughput

- Denote G as the average **traffic load** (in attempts per packet time) of the channel
- The average throughput (in packets per packet time) is

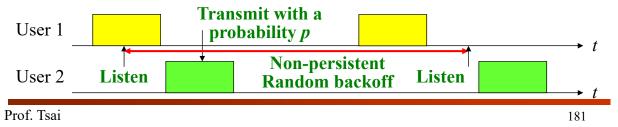
- Pure ALOHA: $S = G \exp[-2G]$, maximum = **0.184** inefficient

- Slotted ALOHA: $S = G \exp[-G]$, maximum = **0.368**



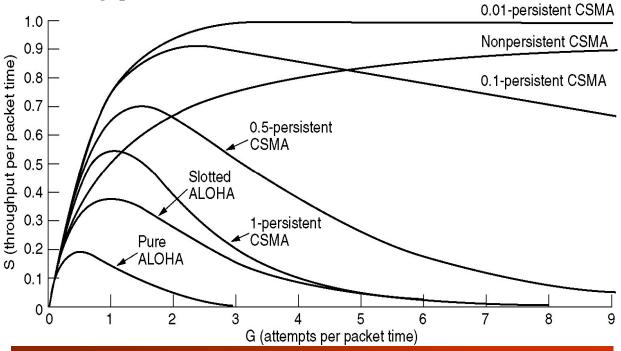
CSMA

- In Carrier Sense Multiple Access (CSMA) scheme, each node listens to a carrier and act accordingly.
- When a node has a data packet to send, it listens to the channel to see if there is anyone else under transmission
 - **p-persistent**: If there is any, it continually senses and waits for the channel to go idle and then transmits with a probability $0 \le p \le 1$, and defers (a random backoff) with probability q = 1 p
 - Non-persistent: If there is any, it defers a random backoff
- When a collision occurs, the user **re-transmits** the packet after a **random backoff time**



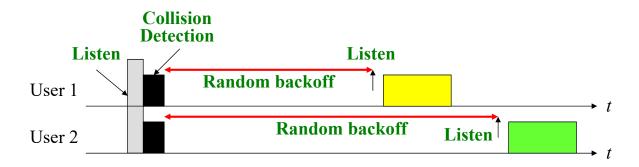
CSMA – Average Throughput

• Throughput versus Traffic load



CSMA/CD

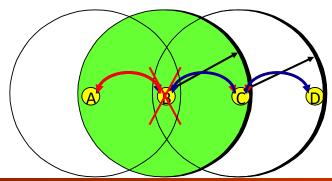
- Carrier Sense Multiple Access with Collision Detection (CSMA/CD): based on CSMA, collision detection is added for performance improvement
- The stations abort their transmission as soon as they detect a collision



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MACA

- In mobile ad hoc networks, random access is more complicated
- Hidden nodes problem: $A \rightarrow B$ are $C \rightarrow B$: collide at B
 - Neither A nor C is aware of this collision
- Exposed node problem: Suppose that B is sending to A
 - Since C can hear B's transmission, it would be a mistake for C to conclude that it cannot transmit to D
 - C's transmission will not interfere with A's receiving



MACA

- 802.11 addresses these problems with an algorithm called Multiple Access with Collision Avoidance (MACA)
 - Before the sender actually transmits any data, the sender and receiver exchange control frames with each other
- The sender transmits a Request to Send (RTS) frame to the receiver
 - Includes a field that indicates **how long** the sender wants to hold the medium
- The receiver replies with a Clear to Send (CTS) frame
 - Echoes the **length field** back to the sender
- Any node that sees the CTS frame knows that it is close to the receiver and cannot transmit for the period of time

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MACA

- There are two more details for MACA
- The receiver sends an **ACK** to the sender after successfully receiving a frame
 - All nodes must wait for this ACK before trying to transmit
- Should two or more nodes detect an idle link and try to transmit an RTS frame at the same time
 - The RTS frames will collide with each other
 - The senders realize the collision has happened when they do not receive the CTS frame after a period of time
 - They wait a random amount of time before trying again
 - Exponential backoff algorithm is generally used

MACA

• RTS/CTS/data/ACK

NAV: network allocation vector DIFS Source RTS Data SIFS SIFS SIFS Destination CTS ACK DIFS NAV (RTS) Other **Contention Window** NAV (CTS) **Defer Access Backoff After Defer**