Example of a **geometrical description** of M-ary QAM: (QAM – OFDM – MIMO)

"From signal waveforms to signal points"

- The **signal space** concept is general and powerful.
- Increased insight and understanding.
- Improved analysis and implementations.
- We can understand more complicated systems.

$$s_{\ell}(t) = A_{\ell}g(t)\cos(2\pi f_{c}t) - B_{\ell}g(t)\sin(2\pi f_{c}t) \qquad \ell = 0, 1, \dots, M - 1 \qquad (2.87)$$

$$s_{\ell}(t) = \underbrace{A_{\ell} \sqrt{E_g/2}}_{s_{\ell,1}} \phi_1(t) + \underbrace{B_{\ell} \sqrt{E_g/2}}_{s_{\ell,2}} \phi_2(t)$$
(2.99)

$$\phi_1(t) = \frac{g(t)\cos(2\pi f_c t)}{\sqrt{E_g/2}}$$
(2.100)

$$\phi_2(t) = -\frac{g(t)\sin(2\pi f_c t)}{\sqrt{E_g/2}}$$
(2.101)

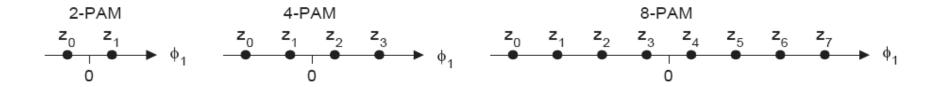
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$$z_{j}(t) = \sum_{\ell=1}^{N} z_{j,\ell} \phi_{\ell}(t) = z_{j,1} \phi_{1}(t) + z_{j,2} \phi_{2}(t) + \dots + z_{j,N} \phi_{N}(t)$$
(5.1)

$$\int_{0}^{T_s} \phi_i(t)\phi_j(t)dt = \begin{cases} 1 & , \quad i=j \\ 0 & , \quad i\neq j \end{cases} \quad i,j=1,2,\dots,N$$
(5.2)

$$z_j(t) \iff \mathbf{z_j} = (z_{j,1}, z_{j,2}, \dots, z_{j,N})^{tr}$$
, $j = 0, 1, \dots, M - 1$ (5.3)

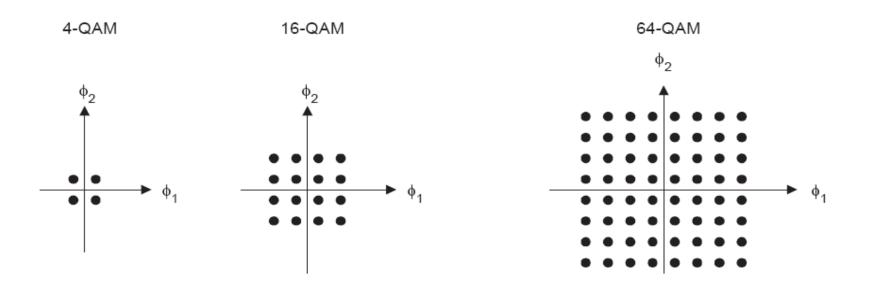
$$s_{\ell}(t) = A_{\ell}g(t) = A_{\ell}\sqrt{E_g} \cdot \underbrace{\frac{g(t)}{\sqrt{E_g}}}_{\phi_1(t)} = \underbrace{A_{\ell}\sqrt{E_g}}_{s_{\ell,1}} \cdot \phi_1(t) = s_{\ell,1} \cdot \phi_1(t)$$
(2.51)



As examples, let us collect some results from subsection 2.4:

M-ary PAM:
$$\boldsymbol{z_j} = ((-M+1+2j)\sqrt{E_g}), \qquad N = 1$$

M-ary PSK: $\boldsymbol{z_j} = \left(\cos(\nu_j)\sqrt{\frac{E_g}{2}}, \sin(\nu_j)\sqrt{\frac{E_g}{2}}\right)^{tr}, \qquad N = 2$
M-ary FSK: $\boldsymbol{z_j} = (0, 0, \dots, \sqrt{E_j}, 0, 0, 0)^{tr}, \qquad N = M$
M-ary QAM: $\boldsymbol{z_j} = \left(A_j\sqrt{\frac{E_g}{2}}, B_j\sqrt{\frac{E_g}{2}}\right)^{tr}, \qquad N = 2$
(5.4)



$$\begin{bmatrix} E_{j} &= \int_{0}^{T_{s}} z_{j}^{2}(t) dt = \sum_{\ell=1}^{N} z_{j,\ell}^{2} = z_{j}^{tr} z_{j} \\ D_{i,j}^{2} &= \int_{0}^{T_{s}} (z_{i}(t) - z_{j}(t))^{2} dt = \sum_{\ell=1}^{N} (z_{i,\ell} - z_{j,\ell})^{2} = \\ &= E_{i} + E_{j} - 2z_{i}^{tr} z_{j} \end{bmatrix}, \quad i,j=0,1,\dots,M-1$$

$$(5.8)$$

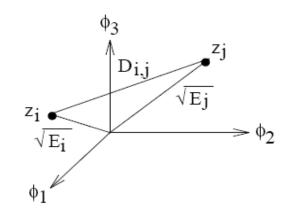
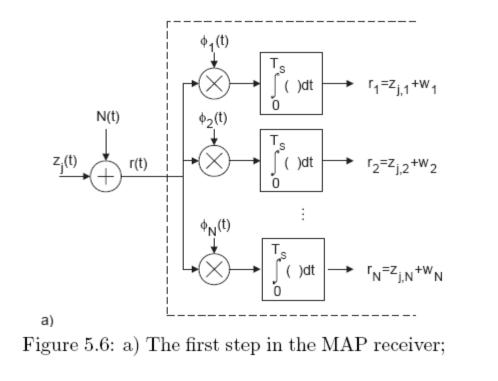


Figure 5.3: Illustrating E_{ℓ} and $D_{i,j}$ in signal space.



$$\int_{0}^{T_{s}} z_{j}(t)\phi_{\ell}(t)dt = \int_{0}^{T_{s}} \sum_{n=1}^{N} z_{j,n}\phi_{n}(t)\phi_{\ell}(t)dt = \sum_{n=1}^{N} z_{j,n} \int_{0}^{T_{s}} \phi_{n}(t)\phi_{\ell}(t)dt = z_{j,\ell}$$
(5.12)

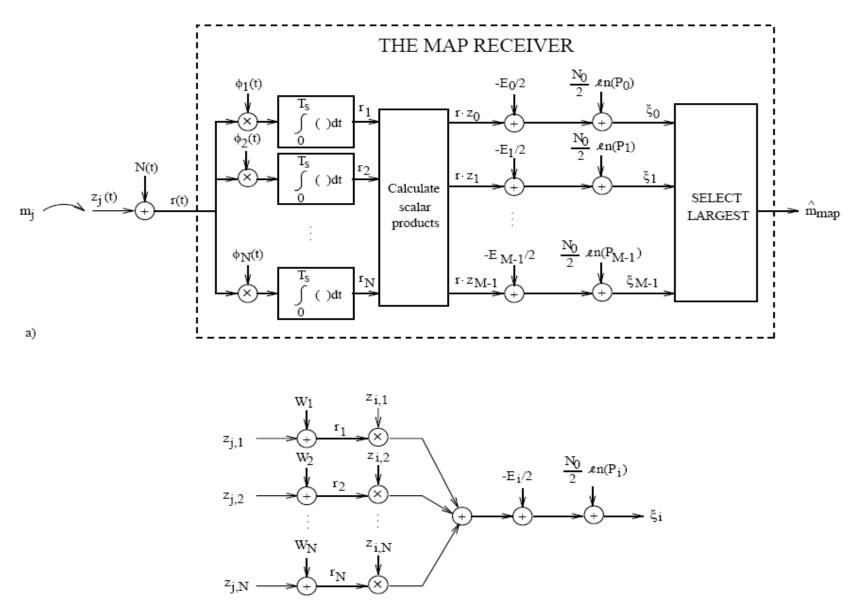
After the correlators we obtain a *received noisy signalpoint* r!

$$E\{w_{\ell}\} = 0$$

$$\sigma_{\ell}^{2} = E\{w_{\ell}^{2}\} = N_{0}/2$$

$$E\{w_{\ell}w_{m}\} = 0, \quad \ell \neq m$$

(5.22)



b)

Figure 5.8: a) The MAP receiver; b) A discrete-time model of the decision variable $\xi_i.$

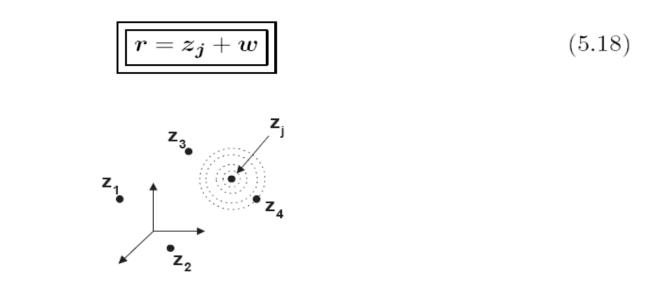


Figure 5.7: Illustrating "the cloud" of noise in r if message m_j is sent.

The distance between the received noisy signal point **r** and the signal point **z**_j is:

$$D_{r,j}^2 = (r - z_j)^{tr} (r - z_j) = \sum_{\ell=1}^N (r_\ell - z_{j,\ell})^2$$
(5.24)

MAP decision rule:

ML decision rule = minimum distance decision rule:

In the MAP decision rule (5.25)–(5.26) we observe that if $P_i = 1/M$, then the terms $N_0 \ell n(P_\ell)$ can be ignored, resulting in the decision rule

$$\hat{m}(r) = m_{\ell} \Leftrightarrow \min_{\{i\}} D_{r,i}^2 = D_{r,\ell}^2$$
(5.28)

Hence, if $P_i = 1/M$, then the ML decision rule is obtained as the minimum Euclidean distance decision rule. Observe also in (5.25) that

5.1.3 The Symbol Error Probability for M-ary PAM

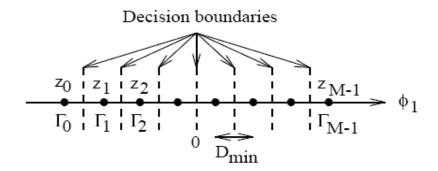


Figure 5.9: The signal space for M-ary PAM with equispaced amplitudes, centered symmetrically around zero (see (5.4)).

$$Prob\{\operatorname{error}|m_0 \operatorname{sent}\} = Prob\left\{w_1 > \frac{D_{\min}}{2}\right\} =$$
$$= Prob\left\{\frac{w_1}{\sqrt{N_0/2}} > \frac{D_{\min}}{\sqrt{2N_0}}\right\} = Q\left(\sqrt{\frac{D_{\min}^2}{2N_0}}\right)$$
(5.31)

$$Prob\{\operatorname{error}|m_{1} \operatorname{sent}\} = Prob\left\{w_{1} < -\frac{D_{\min}}{2} \operatorname{or} w_{1} > \frac{D_{\min}}{2}\right\} =$$
$$= Prob\left\{\frac{w_{1}}{\sqrt{N_{0}/2}} < -\frac{D_{\min}}{\sqrt{2N_{0}}}\right\} + Prob\left\{\frac{w_{1}}{\sqrt{N_{0}/2}} > \frac{D_{\min}}{\sqrt{2N_{0}}}\right\} =$$
$$= 2Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right)$$
(5.32)

$$P_s = \sum_{j=0}^{M-1} P_j Prob\{\text{error}|m_j \text{ sent}\}$$

$$P_s = \frac{2}{M} \left(M - 1 \right) Q \left(\sqrt{\frac{D_{\min}^2}{2N_0}} \right) \,, \quad \text{M-ary PAM}$$
(5.35)

 P_s is shown in Figure 5.13 on page 362.

5.1.4 The Symbol Error Probability for QPSK $r(t) = z_j(t) + N(t), \quad 0 \le t \le T_s, \quad j = 0, 1, ..., M - 1$ (5.13) $r_1 = z_{j,1} + w_1$ (5.36)

$$r_2 = z_{j,2} + w_2 \tag{5.37}$$

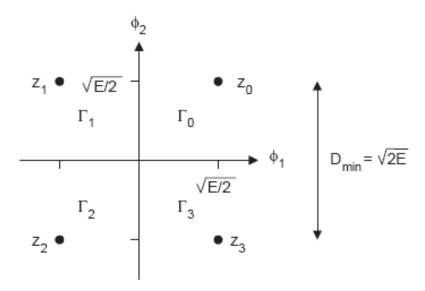


Figure 5.10: The signal space for QPSK if $\nu_{\ell} = (2\pi \frac{\ell}{M} + \pi/4)$ (see (5.4)).

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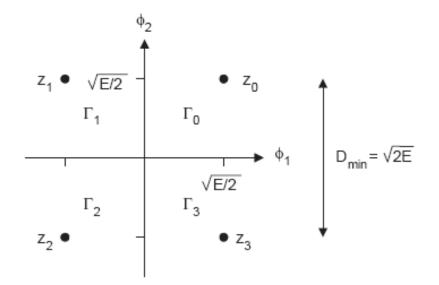


Figure 5.10: The signal space for QPSK if $\nu_{\ell} = (2\pi \frac{\ell}{M} + \pi/4)$ (see (5.4)).

$$Prob\{\operatorname{error}|m_{0} \operatorname{sent}\} = 1 - Prob\{\operatorname{correct decision}|m_{0} \operatorname{sent}\} =$$

$$= 1 - Prob\left\{w_{1} \ge -\frac{D_{\min}}{2}, w_{2} \ge -\frac{D_{\min}}{2}\right\} =$$

$$= 1 - Prob\left\{w_{1} \ge -\frac{D_{\min}}{2}\right\} Prob\left\{w_{2} \ge -\frac{D_{\min}}{2}\right\} =$$

$$= 1 - \left[1 - Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right)\right]^{2} = 2Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right) - Q^{2}\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right) \stackrel{\downarrow}{=}$$

$$= Prob\{\operatorname{error}|m_{j} \operatorname{sent}\}, j = 0, 1, 2, 3$$
(5.38)

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5.1.5 The Symbol Error Probability for M-ary PSK

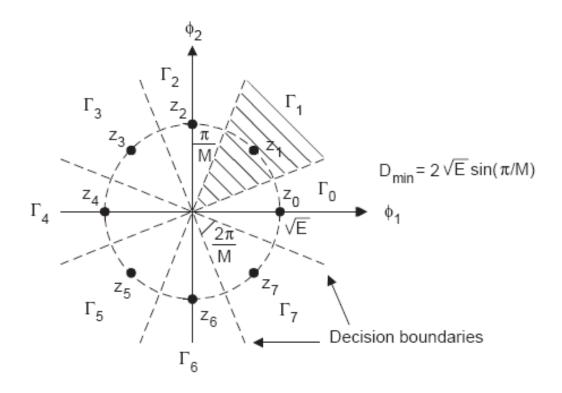


Figure 5.11: The signal space for M-ary PSK if $\nu_{\ell} = 2\pi\ell/M$ (see (5.4)). M = 8 in this figure.

$$\begin{array}{l}
Q\left(\sqrt{\frac{D_{\min}^2}{2N_0}}\right) \leq P_s < 2Q\left(\sqrt{\frac{D_{\min}^2}{2N_0}}\right) \\
D_{\min}^2 = 4E\sin^2(\pi/M)
\end{array}, \text{ M-ary PSK}$$
(5.43)

5.1.6 The Symbol Error Probability for M-ary QAM

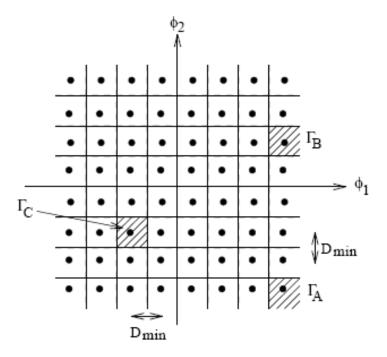


Figure 5.12: The signal space for M-ary QAM (compare with (5.4), see also Subsection 2.4.5.1). M=64 in this figure.

 $\Gamma_A\colon$ Compare with (5.39).

$$Prob\{\text{error}|m_A \text{ sent}\} = 2Q\left(\sqrt{\frac{D_{\min}^2}{2N_0}}\right) - Q^2\left(\sqrt{\frac{D_{\min}^2}{2N_0}}\right)$$
(5.46)

 Γ_B :

$$Prob\{\operatorname{error}|m_{B} \operatorname{sent}\} =$$

$$= 1 - Prob\left\{w_{1} > -\frac{D_{\min}}{2}, -\frac{D_{\min}}{2} \le w_{2} \le \frac{D_{\min}}{2}\right\} =$$

$$= 1 - \left(1 - Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right)\right) \left(1 - 2Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right)\right) =$$

$$= 3Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right) - 2Q^{2}\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right) \qquad (5.47)$$

 Γ_C :

$$Prob\{\operatorname{error}|m_{C} \operatorname{sent}\} =$$

$$= 1 - Prob\left\{-\frac{D_{\min}}{2} \le w_{1} \le \frac{D_{\min}}{2}, -\frac{D_{\min}}{2} \le w_{2} \le \frac{D_{\min}}{2}\right\} =$$

$$= 1 - \left(1 - 2Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right)\right)^{2} =$$

$$= 4Q\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right) - 4Q^{2}\left(\sqrt{\frac{D_{\min}^{2}}{2N_{0}}}\right) \qquad (5.48)$$

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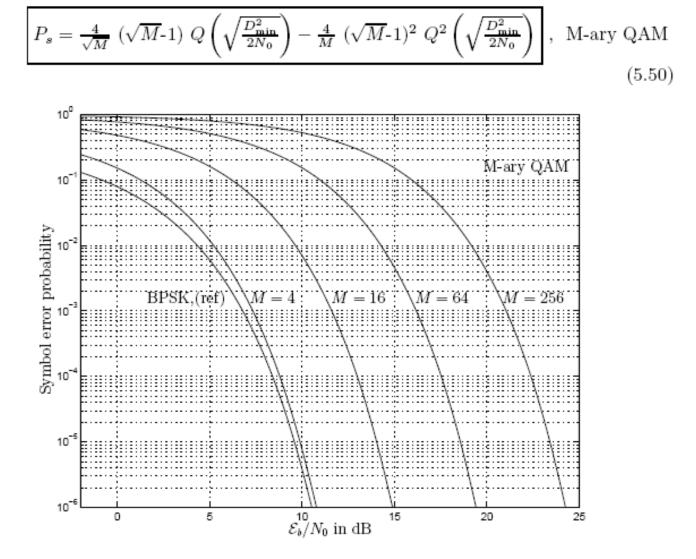


Figure 5.15: The symbol error probability for M-ary QAM, M = 4, 16, 64, 256, see Table 5.1. The specific assumptions are given in Subsection 2.4.5.1 and in Subsection 5.1.6. The bit error probability for BPSK is also given as a reference $(= Q(\sqrt{2\mathcal{E}_b/N_0})).$

5.2.2 Power and Bandwidth Efficiency

We saw in (5.60) that the information bit rate R_b is limited by d_{\min}^2 , c, P_z , N_0 and $P_{s,req}$. Let us divide both sides in (5.60) with the bandwidth W,

$$\rho \leq \frac{d_{\min}^2}{\mathcal{X}} \cdot \frac{\mathcal{P}_z}{N_0 W} = \frac{d_{\min}^2}{\mathcal{X}} \cdot \mathcal{SNR}_r$$
(5.61)

Note that the bandwidth efficiency ρ is limited by d_{\min}^2 , c, $P_{s,req}$, and by the received signal-to-noise power ratio $SNR_r = P_z/N_0W$ within the signal bandwidth W. The bandwidth W is the physical bandwidth defined on the

5.2.3 Shannon's Capacity Theorem

In Shannons capacity theorem, [54], [68], [20], [43], for the bandlimited flat $(|H(f)|^2 = \alpha^2 \text{ within the bandwidth } W)$ AWGN channel, the capacity C for this channel is (in bits per second),

$$\mathcal{C} = W \log_2 \left(1 + \frac{\mathcal{P}_z}{N_0 W} \right) \ , \ [b/s]$$
(5.62)

where W is the physical bandwidth measured on the positive frequency axis containing all the signal power. This remarkable theorem states that ([43], [68]): There exists at least one signal construction method that achieves an arbitrary small error probability, if the bit rate $R_b < C$. If $R_b > C$, then the error probability P_s is high for every possible signal construction method.

$$\begin{bmatrix}
\mathcal{C} = W \log_2 \left(1 + \frac{\mathcal{P}_z}{N_0 W} \right), \quad [b/s]
\end{bmatrix}$$
(5.62)
$$\lim_{W \to \infty} \mathcal{C} = \lim_{W \to} \frac{W}{\ell n(2)} \, \ell n \left(1 + \frac{\mathcal{P}_z}{N_0 W} \right) = \frac{\mathcal{P}_z}{N_0 \ell n(2)}$$
(5.63)
$$\begin{bmatrix}
\frac{\mathcal{C}}{W} = \log_2 \left(1 + \frac{\mathcal{P}_z}{N_0 W} \right) = \log_2 \left(1 + \frac{\mathcal{C}}{W} \cdot \frac{\mathcal{E}_b}{N_0} \right), \quad [bps/Hz]$$
or equivalently,
$$\frac{\mathcal{E}_b}{N_0} = \frac{2^{\mathcal{C}/W} - 1}{\mathcal{C}/W}$$
(5.64)

Since C is the maximum bit rate, \mathcal{E}_b here represents the minimum average received energy per information bit, for a given \mathcal{P}_z , $\mathcal{P}_z = \mathcal{C}\mathcal{E}_b$.

$$\frac{\mathcal{P}_z}{N_0 W} = \frac{\mathcal{C}}{W} \cdot \frac{\mathcal{E}_b}{N_0} = 2^{\mathcal{C}/W} - 1 \tag{5.65}$$

5.2.3.1 Shannon Capacity for General $|H(f)|^2$ and $R_N(f)$

1. For a given average transmitted signal power P_{sent} , and channel quality function $q_{ch}(f) = |H(f)|^2 / R_N(f)$, the parameter B below should first be determined,

$$P_{sent} = \int_{\Omega} \left(B - \frac{R_N(f)}{|H(f)|^2} \right) df$$
(5.68)

This is referred to as "waterfilling"!

2. The capacity C is then found as,

$$C = \int_{\Omega} \frac{1}{2} \log_2 \left(\frac{|H(f)|^2}{R_N(f)} \cdot B \right) df$$
(5.70)

5.4.1 Diversity: Introductory Concepts

"Dont put all eggs in the same basket"

Assume that each message is sent in N dimensions (time/frequency/space etc)

$$s_j(t) = \sum_{n=1}^N s_{j,n} \phi_n(t) , \quad j = 0, 1, \dots, M-1$$
 (5.79)

Assume independent attenuations in each dimension:

$$r(t) = z_j(t) + N(t) = \sum_{n=1}^{N} \alpha_n s_{j,n} \phi_n(t) + N(t)$$

$$z_j = \begin{pmatrix} \alpha_1 & \mathbf{0} \\ \alpha_2 & \\ \mathbf{0} & \ddots & \\ \mathbf{0} & \alpha_N \end{pmatrix} \begin{pmatrix} s_{j,1} \\ \vdots \\ s_{j,N} \end{pmatrix} = \begin{pmatrix} \alpha_1 s_{j,1} \\ \vdots \\ \alpha_N s_{j,N} \end{pmatrix}$$
(5.80)
(5.81)

Note: It can be very "dangerous" to use only one (i.e. N=1) dimension!

We now introduce the concept of **diversity** in connection with Figure 5.21 and (5.80). Diversity is often used, e.g., for so-called **fading** channels (randomly varying signal levels, see Chapter 9), to improve the error probability. *Diversity* can be obtained by spreading the same message over many dimensions. Hence, in the receiver, message m_i has coordinates in, say L, dimensions. Let p denote the probability that a received signal is seriously distorted in any single dimension. The basic idea with diversity is that the probability for large distorsions in **all** dimensions ($\approx p^L$) is significantly lower than p. Observe that this requires that the distorsions in each dimension are essentially independent. So, intuitively speaking, there is a high probability that a few message carrying coordinates "survive" the channel without too much damage, and it is these coordinates that the receiver bases its decision on. Compare with Figure 5.21b,c assuming some of the α_n 's are close to zero. It should also be mentioned here that there is a close relationship between the concept of diversity and the concept of **coding**.

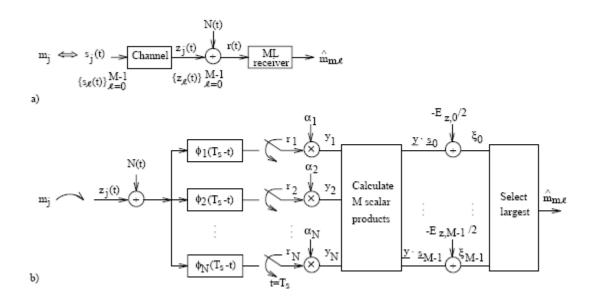


Figure 5.21: a) The digital communication system; b) The ML receiver, assuming (5.80);

Observe that the channel attenuations are used as *multipliers in the receiver* according to the receiver structure in figure 5.8a on page 341!

EXAMPLE 5.23

Assume a binary communication system with equiprobable antipodal signal alternatives,

$$s_1(t) = -s_0(t) = \sum_{k=1}^{K} g_k(t), \quad 0 \le t \le T_b$$

Let $E_{b,sent}$ denote the average transmitted energy per information bit, i.e. $E_{s_1} = E_{s_0} = E_{b,sent}$. It is also assumed that the individual pulses $g_k(t)$ are such that

$$\int_0^{T_b} g_i(t)g_j(t) = \begin{cases} E_{b,sent}/K & , i = j \\ 0 & , i \neq j \end{cases}$$

We can therefore define (sent) basis functions as,

$$\phi_k(t) = \frac{g_k(t)}{\sqrt{E_{b,sent}/K}} , \ k = 1, 2, \dots, K$$

and the signal energy $E_{b,sent}/K$ is sent in each of the K dimensions.

Observe that the situation studied in this example applies to several kinds of diversity, e.g., time- and/or frequency-diversity, depending on how the pulses $g_k(t)$ are chosen.

The communication channel is assumed to be such that the received signal alternatives are,

$$z_1(t) = -z_0(t) = \sum_{k=1}^{K} \alpha_k g_k(t) = \sum_{k=1}^{K} \underbrace{\alpha_k \sqrt{\frac{E_{b,sent}}{K}}}_{z_{1,k}} \phi_k(t)$$

and they are disturbed by AWGN N(t) with power spectral density $R_N(f) = N_0/2$. Note that the channel coefficients $\{\alpha_k\}_{k=1}^K$ multiply the signal in each dimension, respectively. The ideal ML receiver is used and it is assumed that perfect estimates of the channel coefficients are available to the receiver.

- a) Assume that the channel parameters {α_k}^K_{k=1} are known to the receiver. Determine an expression of P_b that includes E_{b,sent}.
- b) Suggest a receiver structure for the case in a).

Solution:

a)

$$P_{b} = Q\left(\sqrt{2\mathcal{E}_{b}/N_{0}}\right)$$

$$\mathcal{E}_{b} = \frac{E_{z_{0}} + E_{z_{1}}}{2} = E_{z_{0}} = E_{z_{1}} = \sum_{k=1}^{K} z_{j,k}^{2} = \frac{E_{b,sent}}{K} \sum_{k=1}^{K} \alpha_{k}^{2}$$

Hence, we obtain that

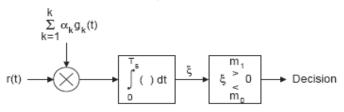
$$P_b = Q \left(\sqrt{\frac{2E_{b,sent}}{N_0 K}} \sum_{k=1}^{K} \alpha_k^2 \right)$$

Note that here a K-fold diversity is obtained, in the sense that signal energy from all K dimensions (or "sub-channels") is efficiently collected and used in the decision process.

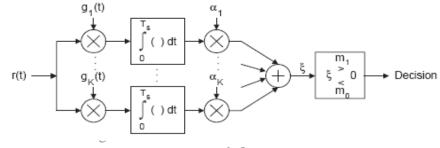
Note also that

$$D_{s_1,s_0}^2 = 4E_{b,sent}$$
$$D_{z_1,z_0}^2 = 4E_z = \frac{4E_{b,sent}}{K} \sum_{k=1}^K \alpha_k^2 = \frac{D_{s_1,s_0}^2}{K} \sum_{k=1}^K \alpha_k^2$$

b) From Figure 4.10 on page 247 we obtain the receiver structure below (the constant 2 is ignored in the correlation below),



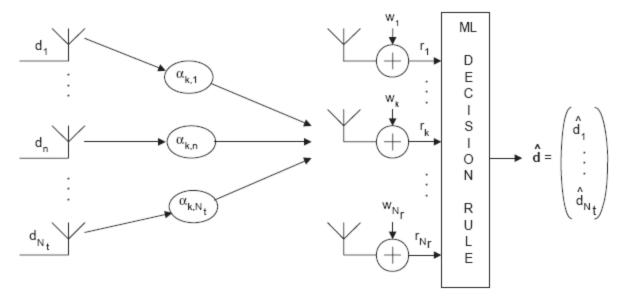
An equivalent receiver structure is also shown below,



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MIMO MODEL



$$r_{k} = \sum_{n=1}^{N_{t}} \alpha_{k,n} d_{n} + w_{k}$$

$$r = \begin{pmatrix} r_{1} \\ \vdots \\ r_{N_{r}} \end{pmatrix} = A \begin{pmatrix} d_{1} \\ \vdots \\ d_{N_{t}} \end{pmatrix} + \begin{pmatrix} w_{1} \\ \vdots \\ w_{N_{r}} \end{pmatrix} = Ad + w$$

Assume, e.g., that: Nt=1 and data symbol d1 is binary:+A or –A

5.4.1.1 An Example Illustrating Diversity Gains

Here we study the case when the channel parameters $\{\alpha_k\}_{k=1}^K$ have the following properties:

- They are assumed to be independent random variables, and only two values are possible for each α_k.
- Each α_k takes the value α_G ("Good") with probability P_G , and the value α_B ("Bad") with probability $P_B = 1 P_G$.

$$\mathcal{E}_{b} = E\left\{\frac{E_{b,sent}}{K}\sum_{k=1}^{K}\alpha_{k}^{2}\right\} = E_{b,sent}E\{\alpha_{k}^{2}\} = E_{b,sent}(\alpha_{G}^{2}P_{G} + \alpha_{B}^{2}(1 - P_{G}))$$
(5.84)

$$P_{b} = E\left\{P_{b|\{\alpha_{k}\}_{k=1}^{K}}\right\} = E\left\{Q\left(\sqrt{\frac{2E_{b,sent}}{N_{0}K}\sum_{k=1}^{K}\alpha_{k}^{2}}\right)\right\} = E\left\{Q\left(\sqrt{\frac{2}{\alpha_{G}^{2}P_{G}+\alpha_{B}^{2}(1-P_{G})}\cdot\frac{\mathcal{E}_{b}}{N_{0}}\cdot\frac{1}{K}\sum_{k=1}^{K}\alpha_{k}^{2}}\right)\right\}$$
(5.85)

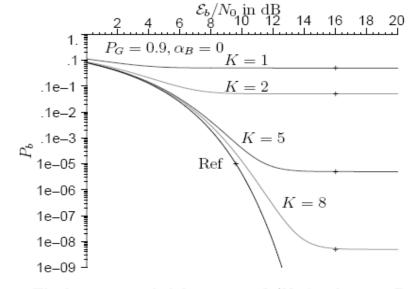


Figure 5.22: The bit error probability versus \mathcal{E}_b/N_0 for the case $P_G = 0.9$ and $\alpha_B = 0$, with K = 1, 2, 5, 8.

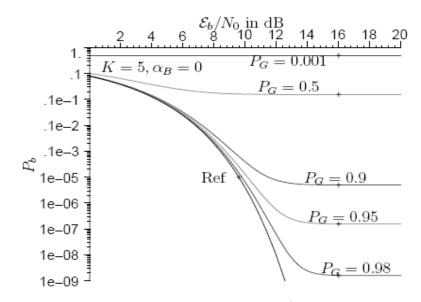


Figure 5.23: The bit error probability versus \mathcal{E}_b/N_0 for the case K = 5 and $\alpha_B = 0$, with $P_G = 0.001, 0.5, 0.9, 0.95, 0.98$.

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3.4.1 Low-Rate QAM-Type of Input Signals

$$x(t) \longrightarrow h(t) \longrightarrow z(t)$$

Figure 3.11: Bandpass filtering.

$$x(t) = x_I(t)\cos(\omega_c t) - x_Q(t)\sin(\omega_c t) = Re\{\tilde{x}(t)e^{j\omega_c t}\}$$
(3.103)

$$\tilde{x}(t) = x_I(t) + jx_Q(t)$$
 (3.104)

This complex signal contains the information!

$$x(t) = x_I(t)\cos(\omega_c t) - x_Q(t)\sin(\omega_c t) = Re\{\tilde{x}(t)e^{j\omega_c t}\}$$
(3.103)

$$z(t) = \int_{-\infty}^{\infty} h(\tau)x(t-\tau)d\tau = \int_{-\infty}^{\infty} h(\tau)Re\{\tilde{x}(t-\tau)e^{j\omega_{c}(t-\tau)}\}d\tau =$$
$$= Re\left\{e^{j\omega_{c}t}\int_{-\infty}^{\infty} h(\tau)\tilde{x}(t-\tau)e^{-j\omega_{c}\tau}d\tau\right\}$$
(3.105)

3 assumptions:

- 1) The duration of the impulse response h(t) can be considered to be equal to T_h . This means that essentially all the energy in h(t) is assumed to be contained within the time interval $0 \le t \le T_h$.
- 2) The input signal is assumed to be a QAM-type of signal with duration $T = T_s$:

$$x(t) = \begin{cases} 0 & , t < 0 \\ A\cos(\omega_c t) - B\sin(\omega_c t) = \sqrt{A^2 + B^2}\cos(\omega_c t + \nu) & , 0 \le t \le T_s \\ 0 & , t > T_s \\ (3.106) \end{cases}$$

3) $T_s > T_h$ ("low" signaling rate).

$$\tilde{x}(t) = \begin{cases} A + jB = \sqrt{A^2 + B^2} e^{j\nu} &, \quad 0 \le t \le T_s \\ 0 &, \quad \text{otherwise} \end{cases}$$
(3.108)

$$T_{h} \leq t \leq T_{s}:$$

$$z(t) = Re \left\{ e^{j\omega_{c}t} \int_{0}^{T_{h}} h(\tau) \sqrt{A^{2} + B^{2}} e^{j\nu} e^{-j\omega_{c}\tau} d\tau \right\} =$$

$$= Re \{ \sqrt{A^{2} + B^{2}} e^{j\nu} \cdot H(f_{c}) e^{j\omega_{c}t} \} =$$

$$= |H(f_{c})| \sqrt{A^{2} + B^{2}} \cos(\omega_{c}t + \nu + \phi(f_{c})) = A_{z} \cos(\omega_{c}t) - B_{z} \sin(\omega_{c}t)$$
(3.109)

Hence, a QAM-signal at the output in this time interval!

However, **attenuation and rotation** compared with the input! Compare with the input x(t) in (3.106)!

$$A_{z} + jB_{z} = (A + jB)H(f_{c}) = \sqrt{A^{2} + B^{2}}|H(f_{c})|e^{j(\nu + \phi(f_{c}))} = = (A + jB)(H_{Re}(f_{c}) + jH_{Im}(f_{c}))$$
(3.110)

$$A_{z} + jB_{z} = (A + jB)H(f_{c}) = \sqrt{A^{2} + B^{2}}|H(f_{c})|e^{j(\nu + \phi(f_{c}))} = = (A + jB)(H_{Re}(f_{c}) + jH_{Im}(f_{c}))$$
(3.110)

A COMPACT MODEL WITH A COMPLEX CHANNEL PARAMETER!!

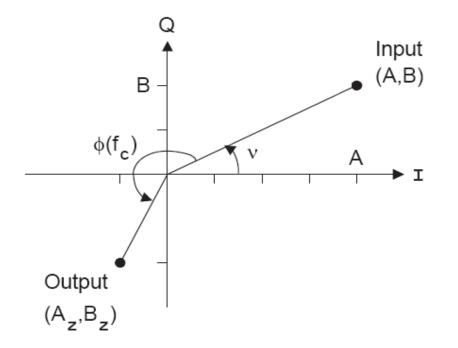


Figure 3.13: Illustrating that the input I-Q amplitudes (A,B) are scaled and rotated by the channel H(f), see (3.109) and (3.110).

$$z(t) = \begin{cases} 0 & , t < 0 \\ \text{"non-stationary transient" starting interval} & , 0 \le t \le T_h \\ |H(f_c)|\sqrt{A^2 + B^2} \cos(\omega_c t + \nu + \phi(f_c)) & , T_h \le t \le T_s \\ \text{"non-stationary transient" ending interval} & , T_s \le t \le T_s + T_h \\ 0 & , t > T_s + T_h \end{cases}$$

and within $T_h \le t \le T_s, A_z + jB_z = (A + jB)H(f_c)$
(3.111)

An important result here is that the input QAM signal x(t) in (3.106) is changed to a new QAM signal by $|H(f_c)|$ and $\phi(f_c)$ in the interval $T_h \leq t \leq T_s$, see also Figure 3.13 and (3.110) how the I-Q components are changed. Furthermore, in OFDM applications the signaling rate $1/T_s$ is low such that $T_s \gg T_h$, and many QAM signals with different carrier frequencies are sent in parallel. Due to linearity, the result in (3.111) can be applied to each QAM signal in the OFDM signal by replacing f_c with f_n . In OFDM applications the receiver uses the time interval $\Delta_h \leq t \leq T_s$ for detection of the output QAM signals, and the duration of this observation interval is denoted $T_{obs} = T_s - \Delta_h$ (compare with (2.110) on page 51, and $T_h \leq \Delta_h$). So, the n:th QAM signal constellation in a sent OFDM signal is attenuated and rotated by H(fn) which is the value of the channel transfer function H(f) at the carrier frequency fn.

3.4.3 N-Ray Channel Model

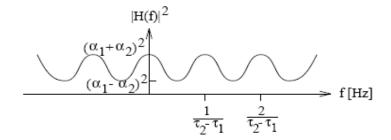
$$z(t) = x(t) * \underbrace{\left(\sum_{i=1}^{N} \alpha_i \delta(t - \tau_i)\right)}_{\text{Impulse response } h(t)} = \sum_{i=1}^{N} \alpha_i x(t - \tau_i)$$
(3.126)

$$H(f) = \mathcal{F}\{h(t)\} = \sum_{i=1}^{N} \alpha_i e^{-j2\pi f\tau_i}$$
(3.128)

So, H(fc) is easy to find!

EXAMPLE 3.20

Rough sketch:



It is seen in this figure that the two signal paths add constructively or destructively (fading) depending on the frequency. Furthermore, if $\alpha_1 \approx \alpha_2$ then |H(f)| is very close to zero at certain frequencies (so-called deep fades)!

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Then (5.135) can be formulated as,

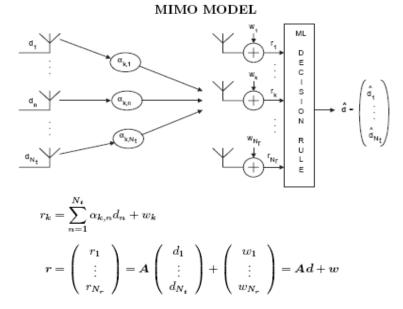
$$r_k = \sum_{n=1}^{N_t} \alpha_{k,n} d_n + w_k \quad , k = 1, 2, \dots, N_r$$
 (5.137)

A compact formulation is now obtained as

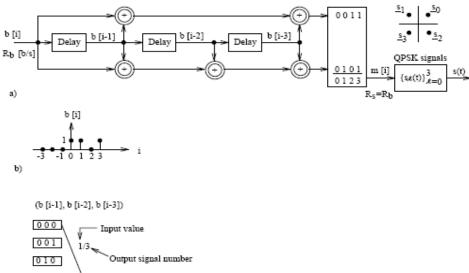
$$r = \begin{pmatrix} r_1 \\ \vdots \\ r_{N_r} \end{pmatrix} = A \begin{pmatrix} d_1 \\ \vdots \\ d_{N_t} \end{pmatrix} + \begin{pmatrix} w_1 \\ \vdots \\ w_{N_r} \end{pmatrix} = Ad + w$$
(5.138)

where the $N_r \times N_t$ matrix A contains the channel coefficients $\{\alpha_{k,n}\}$. The relationship in (5.138) is a basic model in so-called multiple-input multiple-output (MIMO) systems.

The MIMO model is illustrated in the figure below,



64-QAM+Nt=8 (48bits): ML symbol decision rule?



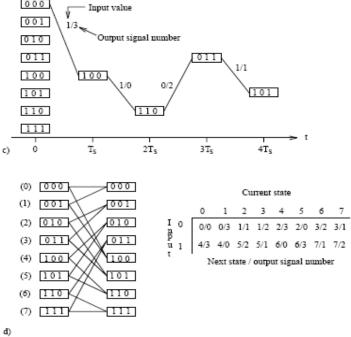
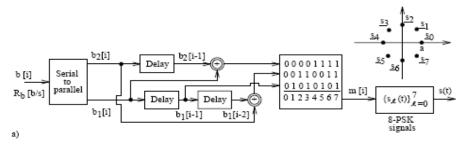
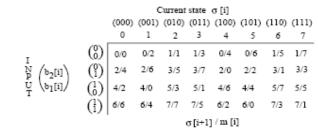


Figure 8.4: a) A rate 1/2 convolutional encoder combined with QPSK signal alternatives; b) A specific input sequence b[i]; c) The corresponding path in the trellis; d) A trellis section, and a table containing all relevant parameters.







b)

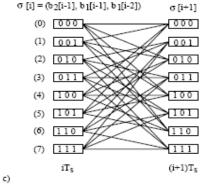


Figure 8.6: a) An example of TCM, from [63]–[64]; b) The mappings $F(\cdot, \cdot)$ and $G(\cdot, \cdot)$; c) A trellis section.

2.32 Let us here study adaptive coding and modulation according to the block diagram below.

$$\mathbf{b} \longrightarrow \text{Encoder} \xrightarrow{\mathbf{c}} \{s_{\ell}(t)\}_{\ell=0}^{\mathsf{M-1}} \longrightarrow \mathbf{s}(t)$$

$$\bar{E}_{sent} = r_c \log_2(M) E_{b,sent} = \frac{k}{n} \log_2(M) E_{b,sent}$$
(8.4)

$$R_{s} = 1/T_{s} = \frac{1}{r_{c}} \cdot \frac{1}{\log_{2}(M)} \cdot R_{b} = \frac{1}{k/n} \cdot \frac{1}{\log_{2}(M)} \cdot R_{b}$$
(8.5)

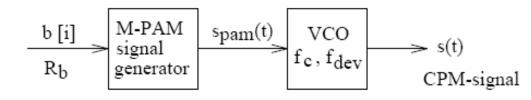
$$W = c \cdot R_s$$
 (8.6)

Typically, the bandwidth W is fixed and given but: the rate of the encoder the number of signal alternatives and the bit rate can be <u>ADAPTIVE</u>, see (8.5)-(8.6)! We have memory in the sequence of sent signal alternatives!

Some sequences are impossible, see problem!

Only "good" sequences are sent!

8.2.1 Continuous Phase Modulation (CPM)



$$s_{pam}(t) = \sum_{n=-\infty}^{\infty} \alpha_n g(t - nT_s), \quad -\infty \le t \le \infty$$
(8.7)

$$s(t) = \sqrt{\frac{2E}{T_s}} \cos(\omega_c t + \theta(t))$$

$$\theta(t) = 2\pi h \sum_{n=-\infty}^{\infty} \alpha_n q(t - nT_s) + \theta_0$$
, CPM (8.20)

Instantaneous frequency ("local frequency"):

$$f_{ins}(t) = f_c + \frac{1}{2\pi} \frac{d\theta(t)}{dt} = f_c + f_{dev} s_{pam}(t) = f_c + f_{dev} \sum_{n = -\infty}^{\infty} \alpha_n g(t - nT_s)$$
(8.12)

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(GSM, Bluetooth)

Examples of pulse shapes:

$$g_{rec}(t) = \begin{cases} 1/2LT_s , & 0 \le t \le LT_s \\ 0 , & \text{otherwise} \end{cases}$$

$$g_{rc}(t) = \begin{cases} [1 - \cos(2\pi t/LT_s)]/2LT_s & 0 \le t \le LT_s \\ 0 , & \text{otherwise} \end{cases}$$

$$(8.9)$$

In these expressions, L is a positive integer which is greater or equal to 1. If L = 1, then the method is referred to as **full response** signaling, and if $L \ge 2$ then we have so-called **partial response** signaling, [2], [43].

Important special cases: *M-ary CPFSK* (continuous phase frequency shift keying) means L=1 and a rectangular pulse shape.

MSK (minimum shift keying) Means h=1/2 + binary CPFSK

$$\theta(t) = 2\pi f_{dev} \sum_{n=-\infty}^{\infty} \alpha_n q(t - nT_s) + \theta_0 \tag{8.15}$$

$$q(t) = \int_0^t g(x) dx$$
 (8.14)

$$q_{rec}(t) = \int_{0}^{t} g_{rec}(x) dx = \begin{cases} 0 & , t \leq 0 \\ t/2LT_{s} & , 0 \leq t \leq LT_{s} \\ 1/2 & , t \geq LT_{s} \end{cases}$$
(8.16)
$$q_{rc}(t) = \int_{0}^{t} g_{rc}(x) dx = \begin{cases} 0 & , t \leq 0 \\ \frac{t}{2LT_{s}} - \frac{1}{4\pi} \sin(2\pi t/LT_{s}) & , 0 \leq t \leq LT_{s} \\ 1/2 & , t \geq LT_{s} \end{cases}$$
(8.17)

$$s(t) = \sqrt{\frac{2E}{T_s}} \cos(\omega_c t + \theta(t))$$

$$\theta(t) = 2\pi h \sum_{n=-\infty}^{\infty} \alpha_n q(t - nT_s) + \theta_0$$
, CPM (8.20)

$$f_{ins}(t) = f_c + \frac{1}{2\pi} \frac{d\theta(t)}{dt} = f_c + f_{dev} s_{pam}(t) = f_c + f_{dev} \sum_{n = -\infty}^{\infty} \alpha_n g(t - nT_s)$$
(8.12)

Instantaneous frequency ("local frequency") within a symbol interval:

$$f_{ins}(t) = f_c + h \sum_{n=-\infty}^{\infty} \alpha_n g(t - nT_s) \stackrel{\downarrow}{=} f_c + \alpha_i \frac{hR_s}{2} = f_c + \alpha_i \frac{hR_b}{2\log_2(M)}, \quad iT_s \le t \le (i+1)T_s \quad (8.22)$$

$$\theta(t) = 2\pi f_{dev} \sum_{n=-\infty}^{\infty} \alpha_n q(t - nT_s) + \theta_0 \tag{8.15}$$

The phase within the i:th symbol interval:

Phase = phase continuity + due to pulseoverlap + due to the current input datasymbol

$$\theta(t) = \pi h \sum_{n=-\infty}^{i-L} \alpha_n + 2\pi h \sum_{\substack{n=i-L+1\\ \text{only if } L \ge 2}}^{i-1} \alpha_n q(t-nT_s) + 2\pi h \alpha_i q(t-iT_s) + \theta_0 ,$$

$$(8.23)$$

The state of the CPM signal:

$$\sigma[i] = \left(\left\{ \pi h \sum_{n=-\infty}^{i-L} \alpha_n \right\}_{\text{mod}2\pi}, \quad \underbrace{\alpha_{i-L+1}, \alpha_{i-L+2}, \dots, \alpha_{i-1}}_{\text{only if } L \ge 2} \right) \right)$$
(8.25)
$$\sigma[i+1] = \left(\left\{ \pi h \sum_{n=-\infty}^{i-L} \alpha_n + \pi h \alpha_{i+1-L} \right\}_{\text{mod}2\pi}, \quad \alpha_{i-L+2}, \alpha_{i-L+3}, \dots, \alpha_i \right)$$
(8.26)

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8.3 ML Reception of Trellis-coded Signals in AWGN

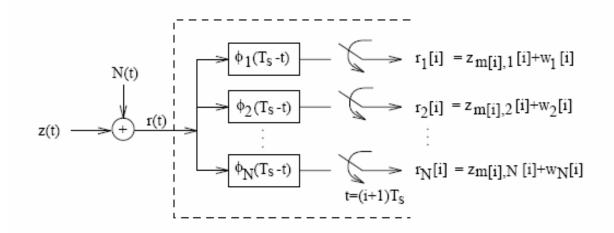


Figure 8.10: The first stage of the ML receiver.

Coherent ML decision rule: "Choose as the decision the "message" corresponding to the signal z(t) that is closest to the received signal r(t) in signal space". Correlator-Outputs:

$$\begin{bmatrix} r[i] = \begin{pmatrix} r_1[i] \\ r_2[i] \\ \vdots \\ r_N[i] \end{pmatrix} = \begin{pmatrix} z_{m[i],1}[i] \\ z_{m[i],2}[i] \\ \vdots \\ z_{m[i],N}[i] \end{pmatrix} + \begin{pmatrix} w_1[i] \\ w_2[i] \\ \vdots \\ w_N[i] \end{pmatrix} = z_{m[i]}[i] + w[i]$$
(8.44)
$$\dots, r[i-1], r[i], r[i+1], \dots$$
(8.45)

ML: The received sequence of noisy signal points should be compared to all possible sequences

$$\dots, \boldsymbol{z}_{m[i-1]}[i-1], \boldsymbol{z}_{m[i]}[i], \boldsymbol{z}_{m[i+1]}[i+1], \dots$$
(8.46)

$$D_{r,z}^{2} = \sum_{n=-\infty}^{\infty} \left(r[n] - z_{m[n]}[n] \right)^{tr} \left(r[n] - z_{m[n]}[n] \right) = \sum_{n=-\infty}^{\infty} \sum_{\ell=1}^{N} \left(r_{\ell}[n] - z_{m[n],\ell}[n] \right)^{2}$$
(8.47)

Up to time t = (i+1)Ts:

Define the accumulated squared Euclidean distance $D_{r,z}^2[i]$ up to time $t = (i+1)T_s$ as,

$$D_{r,z}^{2}[i] = \sum_{n=-\infty}^{i} \left(r[n] - z_{m[n]}[n] \right)^{tr} \left(r[n] - z_{m[n]}[n] \right)$$
(8.48)

Contribution over the i:th symbol interval:

Also define the squared Euclidean distance **increment** at time $t = (i+1)T_s$ as,

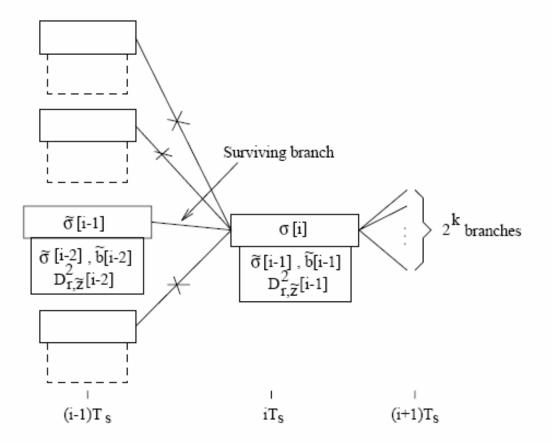
$$D_{inc}^{2}[i] = \left(r[i] - z_{m[i]}[i]\right)^{tr} \left(r[i] - z_{m[i]}[i]\right) = \sum_{\ell=1}^{N} \left(r_{\ell}[i] - z_{m[i],\ell}[i]\right)^{2} \quad (8.49)$$

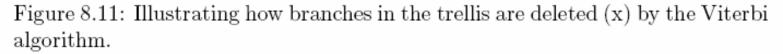
Hence, $D_{inc}^2[i]$ is the squared Euclidean distance contribution obtained in the *i*:th symbol interval $iT_s \leq t \leq (i+1)T_s$.

Recursive calculation:

$$D_{r,z}^{2}[i] = D_{r,z}^{2}[i-1] + D_{inc}^{2}[i]$$
(8.50)

$$D^2_{r, z}[i] = D^2_{r, z}[i-1] + D^2_{inc}[i]$$





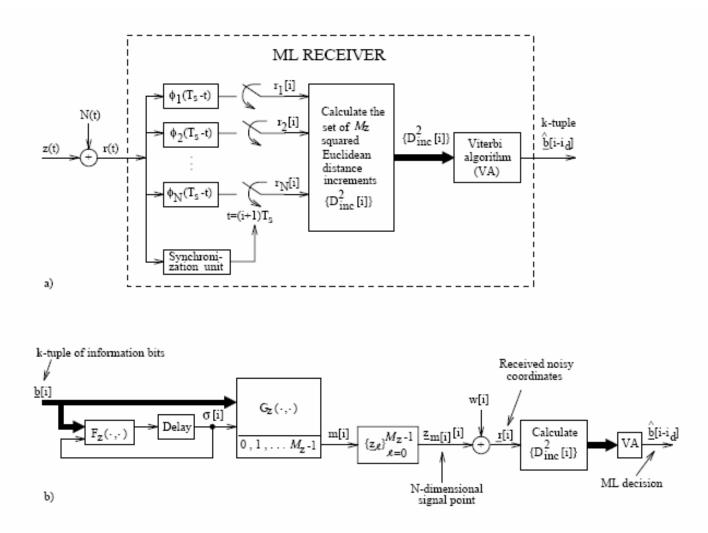
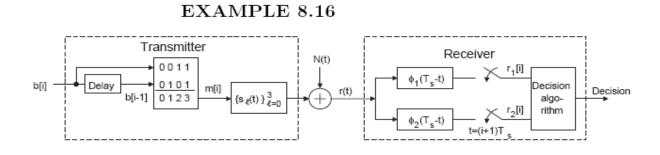
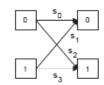


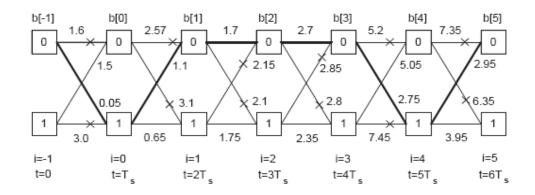
Figure 8.12: a) The coherent ML (sequence) receiver for trellis-coded signals in AWGN; b) A discrete-time model in signal space of the overall digital communication system.



The table below gives the squared Euclidean distance increments $D_{inc}^2[i] = (r_1[i] - s_{j,1})^2 + (r_2[i] - s_{j,2})^2$, obtained between $\mathbf{r}[i]$ and \mathbf{s}_j , j = 0, 1, 2, 3. The noise N(t) is AWGN.

i:	0	1	2	3	4	5
\mathbf{s}_0	1.6	1.07	0.6	1.0	2.5	2.3
$\mathbf{s_1}$	1.5	1.05	1.5	1.1	2.7	0.2
\mathbf{s}_2	0.05	1.6	1.0	1.1	0.05	1.3
83	3.0	0.6	1.1	0.6	5.1	1.2





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8.4 Bit Error Probability for the ML Receiver in AWGN

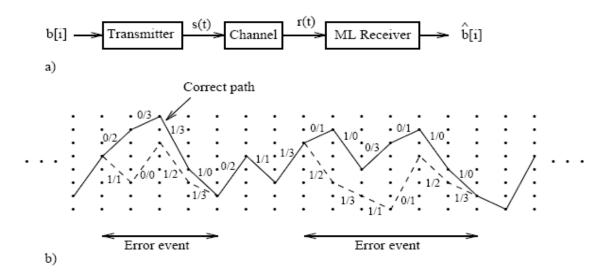


Figure 8.19: a) The digital communication system. b) Parts of the correct path (solid) and the decoded path (dashed) in the trellis.

Observe that by using trellis-coding, longer error events than in the uncoded case are possible. So, with trellis-coding there is a potential for larger Euclidean distances than in the uncoded case.

It can be shown that as a first approximation, at high signal-to-noise ratios, the bit error probability can be approximated with the expression

$$P_b \approx cQ\left(\sqrt{\frac{D_{\min}^2}{2N_0}}\right) = cQ\left(\sqrt{d_{\min}^2 \mathcal{E}_b/N_0}\right)$$
 (8.102)

where D_{\min} is the smallest Euclidean distance in the set of all error events, and $d_{\min}^2 = D_{\min}^2/2\mathcal{E}_b$.

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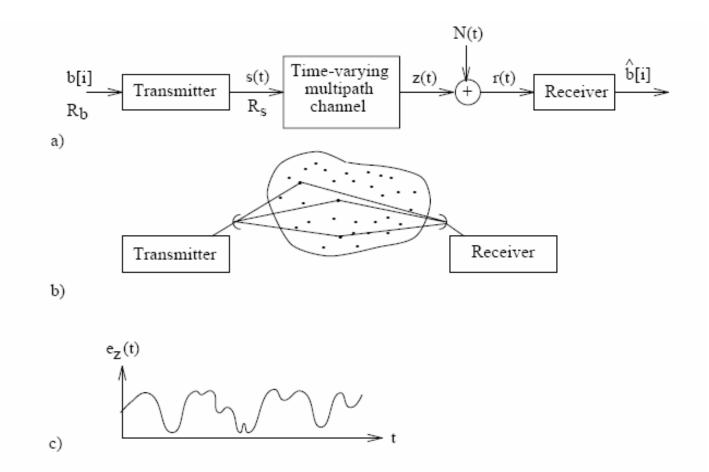


Figure 9.1: a) The digital communication system; b) A scattering medium; c) Illustrating the fading envelope $e_z(t)$.

$$s(t) = \cos((\omega_c + \omega_1)t) , \quad -\infty \le t \le \infty$$
(9.2)

Observe that the quadrature components $z_I(t)$ and $z_Q(t)$ in (9.3) are timevarying. Hence, the output signal z(t) is not a pure sine wave with frequency $f_c + f_1$. This is a significant difference compared with the linear timeinvariant channel. It is seen in (9.3) that the quadrature components depend

9.1.1 Doppler Power Spectrum and Coherence Time

$$R_{\mathcal{D}}(f) = \mathcal{F}(\tilde{c}_{z}(\tau))$$

$$\tilde{c}_{z}(\tau) = \frac{1}{2} E\{[z_{I}(t+\tau) + jz_{Q}(t+\tau)] [z_{I}(t) - jz_{Q}(t)]\}$$

$$R_{z}(f) = \frac{1}{2} (R_{\mathcal{D}}(f+f_{c}+f_{1}) + R_{\mathcal{D}}(f-f_{c}-f_{1}))$$
(9.7)

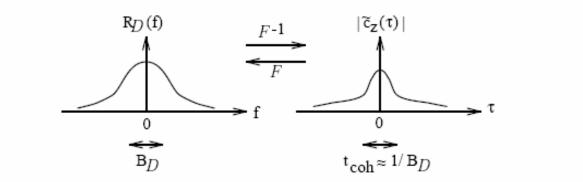


Figure 9.2: Illustrating the Fourier transform pair $\tilde{c}_z(\tau) \longleftrightarrow R_{\mathcal{D}}(f)$.

$$t_{coh} \approx 1/B_{\mathcal{D}} \tag{9.8}$$

9.1.2 Coherence Bandwidth and Multipath Spread

$$z(t) = z(f_1, t) = \underbrace{\frac{1}{2} \quad \tilde{H}_{Re}(f_1, t)}_{z_I(t)} \cos((\omega_c + \omega_1)t) - \underbrace{\frac{1}{2} \quad \tilde{H}_{Im}(f_1, t)}_{z_Q(t)} \sin((\omega_c + \omega_1)t) + \underbrace{\frac{1}{2} \quad \tilde{H}_{Im}(f_1, t)}_{z_Q(t)} \sin((\omega_c + \omega_1)t)$$
(9.9)

What can be said about the output signal z(t) if another frequency $f_2 = f_1 + f_{\Delta}$ is used, instead of f_1 ? Are different frequency-intervals, in the input signal spectrum, treated differently by the time-varying multipath channel? To answer these questions the correlation between $z(f_1, t)$ and $z(f_1 + f_{\Delta}, t)$ can be found by

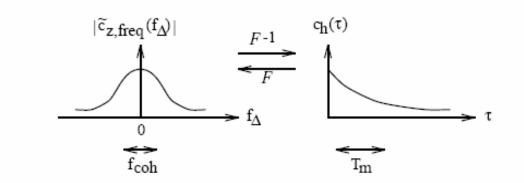


Figure 9.3: Illustrating the Fourier transform pair $c_h(\tau) \longleftrightarrow \tilde{c}_{z,freq}(f_{\Delta})$.

9.2 Frequency-Nonselective, Slowly Fading Channel

$$T_s \ll t_{coh} \tag{9.27}$$

or equivalently,

$$B_{\mathcal{D}} \ll R_s \tag{9.28}$$

This means that the channel is **slowly fading**, which imply that it can be treated as a time-invariant channel within the coherence time.

In this subsection a frequency-nonselective channel is investigated. To obtain this situation it is required that the bandwidth of the transmitted signal, denoted W, is much smaller than the coherence bandwidth f_{coh} of the channel,

$$W \ll f_{coh} \tag{9.29}$$

or equivalently,

$$T_m \ll 1/W \tag{9.30}$$

$$z_{I}(t) + jz_{Q}(t) = \frac{1}{2} (s_{I}(t) + js_{Q}(t))(H_{I} + jH_{Q}) = = e_{s}(t)e^{j\theta_{s}(t)} \cdot ae^{j\phi} = e_{z}(t)e^{j\theta_{z}(t)}$$
(9.37)

$$z(t) = ae_s(t)\cos(\omega_c t + \theta_s(t) + \phi)$$
(9.38)

$$p_a(x) = \frac{2x}{b} e^{-x^2/b}, \quad x \ge 0 \quad (\text{Rayleigh distribution})$$
(9.39)

where,

$$E\{a\} = \frac{1}{2}\sqrt{\pi b}$$
 (9.40)

$$E\{a^2\} = b \tag{9.41}$$

and,

$$p_{\phi}(y) = \begin{cases} 1/2\pi & , -\pi \le y \le \pi \\ 0 & , \text{ otherwise} \end{cases}$$
(9.42)

If we assume uncoded equally likely binary signals over a Rayleigh fading channel $(z_1(t) = as_1(t), z_0(t) = as_0(t))$, then the bit error probability of the ideal

coherent ML receiver is
$$(0 < d^2 = \frac{D_{s_1,s_0}^2}{2E_{b,sent}} \le 2)$$

$$P_b = \int_0^\infty \Pr\{\mathrm{error}|a\} p_a(x) dx = E\{\Pr\{\mathrm{error}|a\}\}$$

(9.43)

$$P_{b} = \int_{0}^{\infty} Q(\sqrt{d^{2}x^{2}E_{b,sent}/N_{0}}) \frac{2x}{b} e^{-x^{2}/b} dx =$$

$$= -e^{-x^{2}/b} Q(x\sqrt{d^{2}E_{b,sent}/N_{0}}) \Big]_{0}^{\infty} - \int_{0}^{\infty} (-e^{-x^{2}/b}) \left(\frac{-\sqrt{d^{2}E_{b,sent}/N_{0}}}{\sqrt{2\pi}} e^{-\frac{x^{2}d^{2}E_{b,sent}/N_{0}}{2}}\right) dx =$$

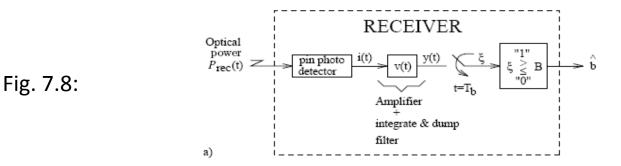
$$= \frac{1}{2} - \sqrt{d^{2}E_{b,sent}/N_{0}} \cdot \beta \underbrace{\int_{0}^{\infty} \frac{e^{-x^{2}/2\beta^{2}}}{\beta\sqrt{2\pi}} dx}_{1/2}$$
(9.44)

$$\mathcal{E}_b = E\{a^2\}E_{b,sent} = bE_{b,sent} \tag{9.45}$$

 $\frac{1}{2d^2\mathcal{E}_b/N_0}$

where $d^2 = 2$ for antipodal signals and $d^2 = 1$ for orthogonal signals. Observe the dramatic increase in P_b due to the Rayleigh fading channel. P_b is no longer exponentially decaying in \mathcal{E}_b/N_0 , it now decays essentially as $(\mathcal{E}_b/N_0)^{-1}$!

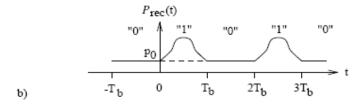
DIVERSITY IS NEEDED!



v(t)

А

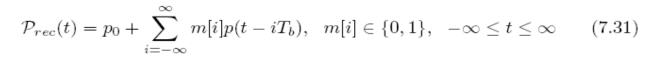
c)



t

Тъ

Received optical power.



$$\xi = y(T_b) = \int_{-\infty}^{\infty} i(\tau)v(T_b - \tau)d\tau = A \int_{0}^{T_b} i(\tau)d\tau = A \int_{0}^{T_b} (i_r(t) + i_d(t))dt = Aq\mathcal{N}_{T_b}$$
(7.32)

q=charge of an electron. id(t)="dark current".

"0": po

"1": po+p(t)

Digital communications - Advanced course:

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Bit error probability:

$$P_b = P_0 \underbrace{\operatorname{Prob}\{\operatorname{error}|m_0 \operatorname{sent}\}}_{P_F} + P_1 \underbrace{\operatorname{Prob}\{\operatorname{error}|m_1 \operatorname{sent}\}}_{P_M}$$
$$= P_0 \operatorname{Prob}\{\xi > B|m_0 \operatorname{sent}\} + P_1 \operatorname{Prob}\{\xi \le B|m_1 \operatorname{sent}\} =$$

$$= P_0 Prob\{\mathcal{N}_{T_b} > (B/Aq)|m_0 \text{ sent}\} +$$

$$+P_1 Prob\{\mathcal{N}_{T_b} \le (B/Aq)|m_1 \text{ sent}\}$$

$$(7.33)$$

$$P_{F} = Prob\{\mathcal{N}_{T_{b}} > \alpha | m_{0} \text{ sent}\} = \sum_{n=\alpha+1}^{\infty} \frac{\mu_{0}^{n} e^{-\mu_{0}}}{n!}$$

$$P_{M} = Prob\{\mathcal{N}_{T_{b}} \le \alpha | m_{1} \text{ sent}\} = \sum_{n=0}^{\alpha} \frac{\mu_{1}^{n} e^{-\mu_{1}}}{n!}$$

$$\alpha = B/Aq$$
(7.35)

We need the averages!

$$\begin{array}{l} P_b \approx Q(\varrho) \\ \varrho = \sqrt{\mu_1} - \sqrt{\mu_0} \end{array}$$

$$\mu_{0} = E\{\mathcal{N}_{T_{b}}|m_{0} \text{ sent}\} = \int_{0}^{T_{b}} \left(\frac{\eta}{hf} p_{0} + \mathcal{I}_{d}\right) dt = \mathcal{I}_{d}T_{b} + \frac{\eta\lambda}{hc} p_{0}T_{b}$$

$$\mu_{1} = E\{\mathcal{N}_{T_{b}}|m_{1} \text{ sent}\} = \mu_{0} + \frac{\eta\lambda}{hc} \int_{0}^{T_{b}} p(t)dt = \mu_{0} + \frac{\eta\lambda}{hc} \cdot \mathcal{E}_{p}$$
(7.34)

$$\mathcal{I}_d = i_d/q$$

$$P_b \approx Q(\varrho)$$
$$\varrho = \sqrt{\mu_1 + \sigma_w^2} - \sqrt{\mu_0 + \sigma_w^2} = \frac{\mu_1 - \mu_0}{\sqrt{\mu_0 + \sigma_w^2} + \sqrt{\mu_1 + \sigma_w^2}}$$

$$\varrho = \frac{\frac{\eta\lambda}{hc} \mathcal{P}_p T_b}{\sqrt{\mathcal{I}_d T_b + \frac{\eta\lambda}{hc} p_0 T_b + k_\sigma T_b} + \sqrt{\mathcal{I}_d T_b + \frac{\eta\lambda}{hc} (p_0 T_b + \mathcal{P}_p T_b) + k_\sigma T_b}} \quad (7.47)$$

$$\mathcal{P}_p = \mathcal{E}_p / T_b$$

$$\frac{\mathcal{P}_{p,1}}{\sqrt{R_{b,1}}} = \frac{\mathcal{P}_{p,2}}{\sqrt{R_{b,2}}} \tag{7.48}$$