

4.17 continued

$$\begin{aligned}
 (s_i - a_i) \cdot (s_j - a_j) &= s_i \cdot s_j - a_i \cdot a_j - a_i \cdot s_j + |a_i|^2 \\
 &= E_p \left\{ \rho + (1-\rho) \delta_{ij} \right\} - \frac{E_p}{M} [2 + (M-1)\rho] + \frac{E_p}{M^2} [2 + (M-1)\rho] \\
 &= E_p \left\{ (1-\rho) \delta_{ij} + \rho \left(1 - \frac{2(M-1)}{M} \right) + \frac{2(M-1)}{M^2} \right\} - \frac{E_p}{M} + \frac{E_p}{M^2} \\
 &= E_p \left\{ (1-\rho) \delta_{ij} - \frac{2}{M} (1-\rho) \right\} \\
 &= \begin{cases} E_p, & i = j \\ -\frac{E_p}{M-1}, & i \neq j \end{cases}
 \end{aligned}$$

Thus, the translated set is a simplex with energy $E_s = E_p(1 - \frac{2}{M})(1 - \rho)$ and hence both the original and the translated set have the same error probability as an M orthogonal signal set with energy $E_0 = E_p(1 - \rho)$. (See p. 261 with E_0 substituted for E_s in Eq. 499.) Note that as expected $\rho = 0 \Rightarrow E_p = E_0$, and $\rho = \frac{1}{M-1} \Rightarrow E_p = E_s$.

5.2

From Eq. 5.15a,

$$\begin{aligned}
 P(\mathcal{E}) &< \exp \left[-X \left(\frac{E_b}{E_{b, \min}} - \ln 2 \right) \right] \\
 &= 2 \left[X \left(\frac{E_b}{E_{b, \min}} - 1 \right) \right] = M \left[\frac{E_b}{E_{b, \min}} - 1 \right]
 \end{aligned}$$

a) We desire $P(\mathcal{E}) < 10^{-6}$. From this bound,

$$1) \frac{E_b}{E_{b, \min}} = 1 \text{ db} = 1.26$$

$$\text{so } M^{-1} = 10^{-6} \text{ or } M = 10^6 / .26 = 10^6 / .26 \approx 10^2 (80 \text{ bits})$$

$$\text{ii) } \frac{E_b}{E_{b, \min}} = 3 \text{ db} = 2$$

$$\text{so } M^{-1} = 10^{-6} \text{ or } M = 10^6 \quad (20 \text{ bits})$$

$$\text{iii) } \frac{E_b}{E_{b, \min}} = 6 \text{ db} = 4$$

$$\text{so } M^{-3} = 10^{-6} \text{ or } M = 100 \quad (7 \text{ bits})$$

$$\text{b) } R = R_T$$

$$R = \frac{1}{.01} = 100 \text{ bits/second}$$

We must, therefore, wait T seconds to build up M messages, where

$$\text{i) } T = \frac{M}{R} = \frac{80}{100} = 800 \text{ msec}$$

$$\text{ii) } \frac{80}{100} = 200 \text{ msec}$$

$$\text{iii) } \frac{100}{100} = 70 \text{ msec}$$

The use of M orthogonal signals in time T says we must provide M dimensions in time T. Thus, D must exceed M/T

$$D = \frac{3}{2} M > M/T$$

so

5.2 continued

$$M > \frac{2M}{3}$$

- i) $M > \frac{2}{3} \frac{10^{24}}{.80} \approx 10^{24}$ ops
- ii) $M > \frac{2}{3} \frac{10^6}{.2} \approx 3.3 \times 10^6$ ops
- iii) $M > \frac{2}{3} \frac{128}{.07} \approx 1.2 \times 10^3$ ops

Note: the bound of Eq. 5.15a is useful to gain insight into bandwidth requirements. The bound of Eq. 5.106 is tighter however. It may be rewritten ($E_{b,\min} \hat{=} \ln 2$)

$$P(\epsilon) < \begin{cases} M \left(\sqrt{\frac{E_b}{E_{b,\min}}} - 1 \right)^2 & 1 < \frac{E_b}{E_{b,\min}} < 4 \\ M \left(\frac{E_b}{2E_{b,\min}} - 1 \right) & 4 < \frac{E_b}{E_{b,\min}} \end{cases}$$

5.3

a) A set of M simplex signals with energy E_s has the same error probability as a set of M orthogonal signals with energy $E_s \frac{M}{M-1}$. Thus

$$P(\epsilon) \leq (M-1) \epsilon \left(\sqrt{\frac{E_s M}{N_0 (M-1)}} \right)$$

$$\leq M e^{-\frac{E_s M}{2N_0 (M-1)}}$$

now

$$E_s = P_s T, \text{ and } M = 2^{RT} = e^{RT \ln 2}$$

$$P(\epsilon)_{\text{simplex}} \leq e^{-T \left[\frac{P_s}{2N_0} \frac{M}{M-1} - R \ln 2 \right]}$$

b) For M biorthogonal signals, we again have, by the symmetry:

$$P(\epsilon)_{\text{biorthogonal}} = P(\epsilon)_{\text{simplex}}$$

The same union bound holds, as always. Two different P_2 values come into consideration, however. The distance from s_1 to every other signal orthogonal to it is obviously

$$d_{\text{ort}}^2 = 2E$$

There is one signal for which $s_k = s_1$, and in this case

$$d_{\text{ent}}^2 = 4E$$

So

$$P(\epsilon|s_1) \leq \sum_{\substack{k=0 \\ i \neq k}}^{M-1} P_2(s_1, s_k) = \epsilon \left(\sqrt{\frac{4E}{2N_0}} \right) + (M-2) \epsilon \left(\sqrt{\frac{2E}{2N_0}} \right)$$

5.2 continued

$$\begin{aligned} < Q(\sqrt{\frac{2E}{2N_0}}) + (M-2)Q(\sqrt{\frac{2E}{2N_0}}) = (M-1)Q(\sqrt{\frac{2E}{2N_0}}) \\ < M Q(\sqrt{\frac{2E}{2N_0}}) \\ < e^{RT \ln 2} e^{-E/2N_0} \\ & RT \ln 2 - \frac{E}{2N_0} = e^{-R \left[\frac{E}{2N_0} - R \ln 2 \right]} \\ & = e \end{aligned}$$

These all have roughly the same error behavior for large M. There are good reasons for the use of each:

- (1) Orthogonal signals are easily generated (for instance sine wave pulse of different frequencies).
- (2) Simplex signals can be generated easily for binary data by a maximal length shift register.
- (3) Biorthogonal signals use only half the bandwidth required by the other two presented here.

5.4

First, let's apply the bandwidth constraint. To do so we need the spectrum. This is

$$X(f) = \int_{-\infty}^{\infty} x(t) e^{-j2\pi f t} dt = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-t^2/2\sigma^2} e^{-j2\pi f t} dt$$

Recalling that $E(e^{jux}) = e^{-1/2(\sigma^2 u^2)}$ for a zero mean Gaussian r.v. x we obtain, with $u = 2\pi f$,

$$X(f) = e^{-1/2(\sigma^2)(2\pi f)^2} = e^{-2\pi^2 \sigma^2 f^2}$$

The shifted pulses add only a linear phase factor but do not affect bandwidth.

Note

$$|X(f)|^2 = e^{-(2\pi\sigma)^2 f^2} = e^{-\left(\frac{f^2}{2(1/8\pi^2\sigma^2)}\right)} = e^{-f^2/2\sigma^2}$$

We desire

$$\frac{\int_{-W}^W |X(f)|^2 df}{\int_{-\infty}^{\infty} |X(f)|^2 df} > .9$$

Note

$\frac{|X(f)|^2}{\int_{-\infty}^{\infty} |X(f)|^2 df}$ is a normalized Gaussian density function

$$W \geq 1.645 \sigma = \frac{1.645}{2\pi\sigma\sqrt{2}}$$

That is, since W is fixed (or selected),

$$\sigma \geq \frac{1.645}{2\pi\sqrt{2}W} \approx .185 \frac{1}{W}$$

Now that we've selected some permissible σ , let's look for a r which meets the overlap condition.

5.4 continued

$$\int_{-\infty}^{\infty} x(t) x(t-\tau) dt \leq .05 \int_{-\infty}^{\infty} x(t) x(t) dt$$

Clearly larger (say 2τ) delays cause less overlap.

However, both integrals can be obtained as the convolution of 2 zero-mean, variance σ^2 Gaussian density functions, one evaluated at τ and the other at 0. The result is a Gaussian density function of zero mean and variance $2\sigma^2$. Hence, the condition on τ

$$\frac{1}{\sqrt{2\pi 2\sigma^2}} e^{-\tau^2/4\sigma^2} \leq .05 \frac{1}{\sqrt{2\pi 2\sigma^2}} e^{-0}$$

or

$$\tau^2/4\sigma^2 > -\ln(.05)$$

$$\tau > 2\sigma \sqrt{-\ln(.05)}$$

$$\approx 3.47\sigma$$

The number of $x(t)$ translates that fit the conditions above and which can be put into T seconds is the integer just below T/τ . As T gets very large, we can say it equals T/τ without making much of an error, percentage-wise.

Thus, the number of dimensions available using these

5.4 continued

pseudo-orthonormal signals is

$$N \approx T/\tau < \frac{T}{3.47\sigma} < \frac{M}{3.47 \times .185} = 1.57 TM$$

that is

$$k \approx 1.57$$

5.6

$$c) \quad \frac{P_B}{N_0} = \frac{JP_B}{N_0} = \begin{cases} P_B \tau & \text{for } A = 2 \\ \frac{P_B \tau}{2} & \text{for } A \geq 3 \end{cases}$$

Then

$$\frac{\overline{E}_N}{N_0} \rightarrow 0$$

is equivalent to

$$\frac{P_B}{N_0} \rightarrow 0 \text{ for } A = 2, \quad \frac{P_B \tau}{2N_0} \rightarrow 0 \text{ for } A \geq 3$$

We need not consider $A = 2$ separately, since it is identical to $A = 4$ for the same value of \overline{E}_N .

Let $X = P_B \tau / N_0 = 2\overline{E}_N$ for $A \geq 3$. Then

$$\lim_{X \rightarrow 0} \left[-\frac{1}{2} \log_2 \left(\frac{1}{A} \sum_{j=1}^A \exp[-X \sin^2 j \frac{\pi}{A}] \right) \right]$$

$$= \lim_{X \rightarrow 0} \left[-\frac{1}{2\sqrt{\pi^2}} \ln \left(\frac{1}{A} \sum_{j=1}^A (1 - X \sin^2 j \frac{\pi}{A}) \right) \right]$$

$$= \lim_{X \rightarrow 0} \left[-\frac{1}{2\sqrt{\pi^2}} \ln \left(1 - \frac{X}{A} \left(\sum_{j=1}^{A-1} \sin^2 j \frac{\pi}{A} + 0 \right) \right) \right]$$

$$= \lim_{X \rightarrow 0} \left[-\frac{1}{2\sqrt{\pi^2}} \ln \left(1 - \frac{X}{2} \right) \right]$$

$$= \frac{X}{4\sqrt{\pi^2}} = \frac{1}{N_0} \frac{1}{2\sqrt{\pi^2}}$$

Thus, for very low rates, A any phase modulated coded signals are as efficient as binary antipodal coded signals. Both can achieve an arbitrarily low error probability if $E_b/N_0 > 2\sqrt{\pi^2}$ and \overline{E}_N/N_0 is sufficiently small. For phase-modulated signals, and in contrast with amplitude-modulated signals for $A > 2$, all M possi-

5.6 continued
ble waveforms have energy $N \overline{E}_N$.

5.6

$$f_k = \sqrt{\frac{2}{A}} \sin 2\pi \left(\frac{t}{T} + \frac{k}{A} \right), -\tau \leq t < 0, \quad k = 1, 2, \dots, A$$

These $f_k(t)$'s can be expanded in two orthonormal functions:

$$f_{k_1}(t) = \sqrt{\frac{2}{A}} \sin 2\pi \frac{t}{T} \quad -\tau \leq t < 0$$

and

$$f_{k_2}(t) = \sqrt{\frac{2}{A}} \cos 2\pi \frac{t}{T} \quad -\tau \leq t < 0$$

for example:



Note that for $A = 2$, this reduces to two binary antipodal signals. From the union bound, Eq. 5.29,

$$\overline{P[\mathbf{E} | m_k]} < \sum_{l=0}^{M_k-1} \sum_{l \neq k} P_2[S_l, S_k]$$

For the white Gaussian noise channel,

$$P_2[S_l, S_k] = Q\left(\frac{|S_l - S_k|}{\sqrt{2N_0}}\right)$$

$$< e^{-|S_l - S_k|^2 / 4N_0}$$

If S_l is specified by the vector $(k_{l_1}, k_{l_2}, \dots, k_{l_y})$ and similarly for S_k , then from (a)

$$|S_l - S_k|^2 = P_0 \tau \sum_{j=1}^y 4 \sin^2 \frac{\pi}{A} (k_{l_j} - k_{k_j})$$

5.6 continued

Over the ensemble of codes, $\frac{1}{A} |k_{1_j} - k_{k_j}|$ is equally likely to have any of the values $0, \frac{1}{A}, \dots, \frac{A-1}{A}$ independently for each $j, j=1, \dots, J$. Hence

$$\begin{aligned} \overline{P_2[S_1, S_k]} &< \sum_{j=1}^J \exp\left\{-\frac{P_0 \tau}{N_0} \sin^2 \frac{\pi}{A} (k_{1_j} - k_{k_j})\right\} \\ &= \prod_{j=1}^J \frac{1}{A} \sum_{l=1}^A \exp\left\{-\frac{P_0 \tau}{N_0} \sin^2 \frac{\pi l}{A}\right\} \\ &= 2^{-NR_0} \end{aligned}$$

in which $(N = 2J$ for $J \geq 3; N = J$ for $J = 2)$

$$R_0 = -\alpha \log_2 \frac{1}{A} \sum_{l=1}^A \exp\left\{-\frac{P_0 \tau}{N_0} \sin^2 \frac{\pi l}{A}\right\}$$

$$\alpha = \begin{cases} 1, & A = 2 \\ 1/2, & A \geq 3 \end{cases}$$

Then,

$$\overline{P[\mathbf{E}]} = \overline{P[\mathbf{E} | m_k]} < M 2^{-NR_0} = 2^{-N(R_0 - R_N)}$$

b) For $A = 2$, we have the case of binary antipodal signaling. For $A = 4$, a code word consists of $N/2$ components from a four-letter biorthogonal alphabet (4 phases differing by 90°) which can be viewed as N components from a binary antipodal alphabet. Hence, R_0 for $A = 2$ must equal R_0 for $A = 4$. For $A = 4$,

$$\begin{aligned} R_0 &= -\frac{1}{2} \log_2 \frac{1}{4} [1 + 2e^{-P_0 \tau / 2N_0} + e^{-P_0 \tau / N_0}] \\ &= -\log_2 \frac{1}{2} [1 + e^{-P_0 \tau / 2N_0}] = -\log_2 \frac{1}{2} [1 + e^{-E_b / N_0}] \end{aligned}$$

Q.E.D.