DIRECT-SEQUENCE SPREAD-SPECTRUM
PERFORMANCE IN A MULTIPATH
FADING ENVIRONMENT

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COPIES OF THIS REPORT SHOULD NOT BE RETURNED UNLESS RETURN IS REQUIRED BY SECURITY CONSIDERATIONS, CONTRACTUAL OBLIGATIONS, OR NOTICE ON A SPECIFIC DOCUMENT.
The performance of coherent direct-sequence spread-spectrum communication systems with specular multipath fading is analyzed in terms of average probability of error. Approximations to such are based on the characteristic-function method. The probability mass function of the correlation receiver output is also derived with this approach. Numerical results are presented for various multipath environments with deterministic gain coefficients and channel coding applications.
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1. Introduction

This thesis examines the performance of coherent phase-shift-keyed (PSK) direct-sequence spread-spectrum (DS/SS) communications in a specular multipath fading environment. In a previous discussion of this particular topic [1], the authors derived an expression for the average probability of error of a correlation receiver, based on the characteristic-function of the multipath interference, given an arbitrary number of paths with both deterministic or random gain coefficients. The channel model considered here is a direct extension of that presented in [1] with some important constraints and modifications. First, there always exists a nonfaded component in the zeroth path signal, i.e., the direct path plus multipath interference with relatively small delay in comparison to the direct path. Second, the range of the path delays is restricted such that intersymbol interference is due to the same pair of adjacent data bits for each of the reflected paths (case(i) in [1]). Third, for the sake of brevity, only deterministic gain coefficients are considered. Finally, the assumption that the path delay must be at least one chip period is removed and consideration is given to delays less than a chip.

As in [1], the average probability of error is approximated using the characteristic-function method. Specifically, the characteristic function of the multipath interference is numerically integrated using Simpson’s method. This technique is extended also to compute the probability mass function of the receiver output.
The thesis is presented in the following fashion. Section 2 discusses the system and channel models. Section 3 provides a derivational analysis of the average probability of error for any desired range of path delays within model constraints. In Section 4, an analysis is provided for the performance of channel coding applications. Then, the models are numerically evaluated for the PSK DS/SS system in Section 5. Finally, the conclusions of these results are presented in Section 6. Following the text are two appendices. Appendix A links the characteristic-function method to an expression for the average probability of error. Appendix B contains the computer code used to produce the numerical results.
2. System and Channel Models

Transmitter

As presented in [1], a channel user starts with a data signal \( b(t) \), a sequence of rectangular pulses, each of duration \( T \), where the \( l \)-th pulse has amplitude \( b_l \) for \( lT < t < (l+1)T \). The data \( (b_l) \) is viewed as a sequence of mutually independent random variables, each with equal probability of being +1 or -1. The signal is then spread by the waveform \( x(t) \), given by

\[
x(t) = x_i \nu(t - iT_z), \quad iT_z \leq t < (i + 1)T_z
\]

where \( x_i \) is the binary signature sequence and \( \nu(s) \) is a rectangular chip waveform satisfying \( \nu(s) = 1 \) for \( 0 \leq s < T_z \) and \( \nu(s) = 0 \) otherwise [2].

\( T_z \) represents the chip duration for the transmitted signal. The processing gain \( N \) is defined as the ratio between the period of the data signal and the chipping period: i.e., \( N = T/T_z \). Thus, the signature sequence \( (x_i) \) has a period of \( N \) also. Considering the transmitter power \( P \) and the phase angle of the PSK modulator \( \theta \), the created spread-spectrum signal \( u(t) \) can be expressed as

\[
u(t) = \sqrt{2P}b(t)x(t)\exp(j\theta).
\]

Thus, with carrier frequency \( f_c \), the transmitted signal for a PSK DS/SS system is given by

\[
s(t) = Re[u(t)\exp(j2\pi f_c t)].
\]
Channel

The output signal \( y(t) \) produced by the specular multipath channel consists of one direct path and \( L \) delayed paths and may be modeled as the sum of attenuated, phase-shifted, and delayed versions of the input signal \( s(t) \). Thus, as in [1], it may be expressed as

\[
y(t) = \sum_{n=0}^{L} \text{Re}\{g_n u(t - \tau_n) \exp[j2\pi f_c(t - \tau_n)]\}
\]

(4)

where the complex gain coefficients \( g_n \) are given as

\[
g_0 = 3 + \gamma_0 A_0 \exp(j\theta_0) \quad \text{and} \quad g_n = \gamma_n A_n \exp(j\theta_n), \quad 1 \leq n \leq L
\]

(5)

for the \( n \)th delayed path. Here, \( \gamma_n A_n \) and \( \theta_n \) represent the attenuation and phase shift incurred from the reflection of the delayed paths, respectively, while \( \tau_n \) is the \( n \)th path delay relative to the zeroth, or direct path, assuming that \( \tau_0 = 0 \). The parameter \( 3 \) signifies the gain coefficient for the nonfaded component of the direct path. Throughout the analysis, it is assumed that \( 3 = 1 \) and the random variables \( A_n \) are normalized to have second-order moments equal to 1. Also, \( \gamma_n^2 \) may be interpreted as the relative power of the \( n \)th delayed path as compared to the nonfaded component of the direct path.

As previously mentioned, the zeroth channel consists of a direct path plus the combination of all multipath interference with small relative delays in comparison. The \( n \)th component channel \((n > 0)\) then represents multipath interference combinations with relative delay of approximately
The variables $A_n$, $\theta_n$, and $\tau_n$ are modeled as mutually independent random vectors with mutually independent components as well. That is, $A = (A_1, A_2, \ldots, A_L)$, $\theta = (\theta_1, \theta_2, \ldots, \theta_L)$, and $\tau = (\tau_1, \tau_2, \ldots, \tau_L)$. It is also assumed that these quantities are independent of $A_0$ and $\theta_0$.

In general, the range of the path delays is $0 < \tau_n < T$. To allow resolvable demodulation of the transmitted signal, this range is normally limited such that $T_1 \leq \tau_n \leq T - T_2$. That is, the delays are at least one chipping period away from zero, to provide necessary receiver rejection of the delayed paths, and at least one chipping period away from $T$, so that the receiver will not demodulate the delayed paths as direct paths with one-period delay. However, for analytical purposes, this paper considers situations in which $0 \leq \tau_n \leq T - T_c$; the upper limit is preserved to model practical applications.

**Receiver**

The output of the channel $y(t)$ is combined with additive white Gaussian noise $n(t) \sim N(0, \frac{N_0}{2})$, producing the received signal $r(t)$. The receiver, in turn, correlates this signal with $\sqrt{2/T} x(t) \cos(2\pi f_c t)$ and integrates from 0 to $T$. Thus, the receiver output becomes

$$Z = \sqrt{\frac{2}{T}} \int_0^T r(t)x(t)\cos(2\pi f_c t)dt. \quad (6)$$

Making the necessary substitutions, (6) becomes
\[ Z = \sqrt{\frac{2}{T}} \left\{ \eta_0 + \sqrt{2P} \left( \int_0^T b(t)x^2(t) [3 \cos(2\pi f_c t) + \gamma_0 A_0 \cos(2\pi f_c t + \theta_0)] \cos(2\pi f_c t) dt \
+ \sum_{n=1}^L \int_0^T \gamma_n A_n b(t - \tau_n)x(t - \tau_n)x(t) \cos[2\pi f_c (t - \tau_n) + \theta_n] \cos(2\pi f_c t) dt \right) \right\} \]

where \( \eta_0 \sim N(0, \frac{N_0 T}{4}) \). Because \( f_c \) is much larger than \( 1/T \) in practical systems, the double-frequency terms above may be ignored. Thus, the integrated output becomes

\[ Z = \sqrt{\frac{2}{T}} \left\{ \eta_0 + \frac{\sqrt{P}}{2} \left[ b_0 (3 + \gamma_0 A_0 \cos(\theta_0)) \int_0^T dt \
+ \sum_{n=1}^L \int_0^T \gamma_n A_n b(t - \tau_n)x(t - \tau_n)x(t) \cos(\phi_n) dt \right] \right\} \quad (7) \]

where \( \phi_n = \theta_n - 2\pi f_c \tau_n \). Both \( \theta_n \) and \( \theta_0 \) are assumed to be uniformly distributed in \([0, 2\pi]\). Similarly, because of the size of \( f_c \) in comparison to \( \tau_n \), the modulo-2\( \pi \) form for \( \phi_n \sim U[0, 2\pi] \). Defining the multipath interference component for the zeroth path \( F_0 = A_0 b_0 \cos(\theta_0) \) and splitting the range of the integrand into the ranges \([0, \tau_n]\) and \([\tau_n, T]\), (7) becomes

\[ Z = \sqrt{\frac{2}{T}} \left\{ \eta_0 + \frac{\sqrt{P}}{2} \left[ (3b_0 + \gamma_0 F_0)T + \sum_{n=1}^L \gamma_n A_n \cos(\phi_n) \left( \int_0^{\tau_n} b(t - \tau_n)x(t - \tau_n)x(t) dt 
+ \int_{\tau_n}^T b(t - \tau_n)x(t - \tau_n)x(t) dt \right) \right] \right\}. \]

From [3], the continuous partial autocorrelation functions of \( x(t) \) are defined by
\[ R_x(\tau_n) = \int_0^{\tau_n} x(t - \tau_n)x(t)dt. \quad 0 \leq \tau_n \leq T \quad \text{and} \quad \hat{R}_x(\tau_n) = R_x(T - \tau_n). \]

Thus, making these substitutions,

\[ Z = \sqrt{\frac{2}{T}} \left\{ \eta_0 + \sqrt{\frac{P}{2}} \left[ (\beta b_0 + \gamma_0 F_0)T + \sum_{n=1}^{L} \gamma_n A_n \cos(\phi_n) \left( b_{-1} R_x(\tau_n) + b_0 \hat{R}_x(\tau_n) \right) \right] \right\}. \quad (8) \]

The \( L \) other multipath interference components are given by

\[ F_n = T^{-1} A_n [b_{-1} R_x(\tau_n) + b_0 \hat{R}_x(\tau_n)] \cos(\phi_n). \quad (9) \]

Thus, it follows from (8).

\[ Z = \sqrt{\frac{2}{T}} \left\{ \eta_0 + \sqrt{\frac{P}{2}} \left[ (\beta b_0 + \gamma_0 F_0)T + \sum_{n=1}^{L} \gamma_n F_n T \right] \right\}. \]

Finally, defining \( E_b = PT \) as the energy per information bit of the transmitted signal, the integrated output reduces to

\[ Z = \eta_0' + \sqrt{E_b} \left( \beta b_0 + \sum_{n=0}^{L} \gamma_n F_n \right) \quad (10) \]

where \( \eta_0' \sim N(0, \frac{N_0}{2}) \).
3. Average Probability of Error

The method applied in evaluating the average probability of error for the DS/SS system is the characteristic-function method, as employed in [1]. Here, the approximation is based on the integration of the characteristic-functions of the noise and multipath interference components of the correlated receiver. The analysis which follows considers deterministic gain coefficients $g_n$.

The gain coefficients are normalized such that $\beta = 1$ and $A_n$ does actually equal 1 for $0 \leq n \leq L$. From (10), define the following:

$$\eta = \eta_0 / \sqrt{E_b} \quad (11)$$

$$I = \sum_{n=0}^{L} \gamma_n F_n. \quad (12)$$

This, in turn, implies that

$$Z = \sqrt{E_b}(\eta + b_0 + I). \quad (13)$$

Since the decision of the receiver is based on comparing the output $Z$ with a zero threshold, the average probability of error conditioned on $b_0$ is written as

$$\tilde{P}_e = \frac{1}{2} P(Z \leq 0 | b_0 = +1) + \frac{1}{2} P(Z > 0 | b_0 = -1)$$

$$= \frac{1}{2} [P(\eta + I \leq -1) + P(\eta + I > 1)]$$

$$= \frac{1}{2} - \frac{1}{2} P(-1 < \eta + I \leq 1). \quad (14)$$
As demonstrated in Appendix A, this expression becomes

\[ \hat{P}_e = Q(\sigma^{-1}) + \frac{1}{\pi} \int_0^\infty \text{sinc}(u/\pi) \phi_n(u)[1 - \phi_I(u)] du \]  

(15)

where \( Q(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty \exp(-t^2/2) dt \), \( \sigma^2 = (2E_b/N_0)^{-1} \) is the variance of \( \eta \).

and \( \phi_n \) and \( \phi_I \) are the respective characteristic functions of \( \eta \) and \( I \). Since \( \eta \) is a Gaussian random variable, its characteristic function is given by

\[ \phi_n(u) = \exp \left( iu\bar{\eta} - \frac{1}{2}u^2\sigma^2 \right) = \exp \left( -\frac{1}{2}u^2\sigma^2 \right) \]  

(16)

recalling that \( \eta \) is zero mean. The characteristic function of the interference sum \( I \), can be expressed as the product

\[ \phi_I = \tilde{\phi}_0(u)\tilde{\phi}(u) \]  

(17)

where \( \tilde{\phi}_0(u) \) represents the characteristic function of the direct-path faded component \( \gamma_0F_0 \) and \( \tilde{\phi}(u) \) denotes the characteristic function for the interference components \( \sum_{n=1}^L \gamma_nF_n \). For the direct-path,

\[ \tilde{\phi}_0(u) = E[\exp(ju\gamma_0F_0)] = \frac{1}{2\pi} \int_0^{2\pi} \exp(ju\gamma_0b_0 \cos \theta) d\theta = \frac{1}{2\pi} \int_0^{2\pi} \cos(u\gamma_0b_0 \cos \theta) d\theta = \frac{2}{\pi} \int_0^{\pi/2} \cos(\gamma_0u \cos \theta) d\theta \]  

(18)

where the sine component of the complex exponential integrates to zero.

From (10), there are two cases to consider in evaluating \( \tilde{\phi}(u) \): \( b_1 = b_0 \) and
$b_{-1} \neq b_0$, both of which are equally likely to occur. Also, as opposed to the path delay restriction of [1], $\tau$ is assumed to be uniformly distributed in $[M_1 T_c, M_2 T_c]$, where $M_1$ and $M_2$ are integers and $M_1 < M_2$. Thus,

$$\tilde{\phi}(u) = \frac{1}{2} E \left[ \exp \left( ju \sum_{n=1}^{L} \gamma_n F_n(b_{-1} = b_0) \right) \right] + \frac{1}{2} E \left[ \exp \left( ju \sum_{n=1}^{L} \gamma_n F_n(b_{-1} \neq b_0) \right) \right]$$

$$= \frac{1}{2} \prod_{n=1}^{L} E \left[ \exp \left( ju \gamma_n F_n(b_{-1} = b_0) \right) \right] + \frac{1}{2} \prod_{n=1}^{L} E \left[ \exp \left( ju \gamma_n F_n(b_{-1} \neq b_0) \right) \right]$$

$$= \frac{1}{2} \prod_{n=1}^{L} M \int_{0}^{2\pi} \int_{M_1 T_c}^{M_2 T_c} \exp \left[ ju \gamma_n \left( R_x(\tau) + \hat{R}_x(\tau) \right) \frac{\cos \phi}{T} \right] d\tau d\phi$$

$$+ \frac{1}{2} \prod_{n=1}^{L} M \int_{0}^{2\pi} \int_{M_1 T_c}^{M_2 T_c} \exp \left[ ju \gamma_n \left( -R_x(\tau) + \hat{R}_x(\tau) \right) \frac{\cos \phi}{T} \right] d\tau d\phi$$

(19)

where the constant $M$ is defined by

$$M = \frac{1}{2\pi (M_2 - M_1) T_c}.$$

Because $I$ has symmetric distribution, the characteristic functions $\tilde{\phi}_0$ and $\tilde{\phi}$ must be real and even; i.e., $\tilde{\phi}(u) = \tilde{\phi}(-u)$. The autocorrelation functions $R_x$ and $\hat{R}_x$ are defined in [3] in terms of the discrete aperiodic autocorrelation function $C_x$ of the signature sequence $(x_i)$, and the chip waveform partial autocorrelation functions $R_\psi$ and $\hat{R}_\psi$. That is,

$$R_x(\tau) = C_x(N-l)\hat{R}_\psi(\tau - lT_c) + C_x(N-l-1)R_\psi(\tau - lT_c)$$

and

$$\hat{R}_x(\tau) = C_x(l)\hat{R}_\psi(\tau - lT_c) + C_x(l+1)R_\psi(\tau - lT_c)$$
where \( l \) denotes the integer part of \( \tau/T_c \). \( \hat{R}_c(s) = \int_{s}^{s+T_c} \psi(t)\psi(t-s)dt \) and
\( R_\psi(s) = \int_{s}^{s+T_c} \psi(t)\psi(t+T_c-s)dt \) for \( 0 \leq s \leq T_c \). Thus, (19) becomes

\[
\hat{\hat{\phi}}(u) = \frac{1}{2} \prod_{n=1}^{L} \int_{0}^{2\pi} \int_{M_1T_c}^{M_2T_c} \cos \{ \varphi_{n}(u) \cos [(C_x(N) + C_x(l + 1))R_\psi(\tau - lT_c) - (C_x(N) - C_x(l + 1))R_\psi(\tau - lT_c)] \} d\tau d\phi
\]

\[
+ \left( C_x(N) - C_x(l + 1) \right) \hat{R}_\psi(\tau - lT_c) \cos \frac{\phi}{T} \right \} d\tau d\phi \]

\[
+ \frac{1}{2} \prod_{n=1}^{L} \int_{0}^{2\pi} \int_{M_1T_c}^{M_2T_c} \cos \{ \varphi_{n}\left( -C_x(N) - C_x(l + 1) \right)R_\psi(\tau - lT_c) \} \cos \frac{\phi}{T} \}
\]

\[
+ \left( -C_x(N) - C_x(l + 1) \right) \hat{R}_\psi(\tau - lT_c) \cos \frac{\phi}{T} \right \} d\tau d\phi \]

From [4], the periodic (even) and odd autocorrelation functions for the chip waveform are respectively defined by

\[
\theta_x(l) = C_x(l) + C_x(l - N), \quad 0 \leq l \leq N - 1
\]

and

\[
\hat{\theta}_x(l) = C_x(l) - C_x(l - N), \quad 0 \leq l \leq N - 1.
\]

Making these substitutions,

\[
\hat{\phi}(u) = \frac{1}{2} \prod_{n=1}^{L} \int_{0}^{2\pi} \int_{M_1T_c}^{M_2T_c} \cos \left\{ u\varphi_{n} \left[ \theta_x(l + 1)R_\psi(\tau - lT_c) + \theta_x(l)\hat{R}_\psi(\tau - lT_c) \right] \frac{\cos \phi}{T} \right \} d\tau d\phi
\]

\[
+ \frac{1}{2} \prod_{n=1}^{L} \int_{0}^{2\pi} \int_{M_1T_c}^{M_2T_c} \cos \left\{ u\varphi_{n} \left[ \hat{\theta}_x(l + 1)R_\psi(\tau - lT_c) + \hat{\theta}_x(l)\hat{R}_\psi(\tau - lT_c) \right] \frac{\cos \phi}{T} \right \} d\tau d\phi
\]

for \( 0 \leq l \leq N - 2 \). Considering the variable substitution \( s = \tau - lT_c \),
\[ o(u) = \frac{1}{2} \prod_{n=1}^{L} M \int_{0}^{2\pi} \sum_{l=M_1}^{M_2-1} \int_{0}^{T_c} \cos \left\{ u \gamma_n \left[ \theta_x(l+1)R_v(s) + \theta_x(l)\hat{R}_v(s) \right] \frac{\cos \phi}{T} \right\} dsd\phi \]

\[ + \frac{1}{2} \prod_{n=1}^{L} M \int_{0}^{2\pi} \sum_{l=M_1}^{M_2-1} \int_{0}^{T_c} \cos \left\{ u \gamma_n \left[ \hat{\theta}_x(l+1)R_v(s) + \hat{\theta}_x(l)\hat{R}_v(s) \right] \frac{\cos \phi}{T} \right\} dsd\phi \]

\[ = \frac{1}{2} \prod_{n=1}^{L} 4M \sum_{l=M_1}^{M_2-1} \int_{0}^{\pi/2} \int_{0}^{T_c} \cos \left\{ u \gamma_n \left[ \theta_x(l+1)R_v(s) + \theta_x(l)\hat{R}_v(s) \right] \frac{\cos \phi}{T} \right\} dsd\phi \]

\[ + \frac{1}{2} \prod_{n=1}^{L} 4M \sum_{l=M_1}^{M_2-1} \int_{0}^{\pi/2} \int_{0}^{T_c} \cos \left\{ u \gamma_n \left[ \hat{\theta}_x(l+1)R_v(s) + \hat{\theta}_x(l)\hat{R}_v(s) \right] \frac{\cos \phi}{T} \right\} dsd\phi \]

\[ = \frac{1}{2} \left\{ \prod_{n=1}^{L} \left[ \frac{1}{M_2 - M_1} \sum_{l=M_1}^{M_2-1} f(\gamma_n u; l, \theta_x) \right] \right\} \] (20)

where

\[ f(v; l, h) = \frac{2}{\pi T_c} \int_{0}^{\pi/2} \int_{0}^{T_c} \cos \left\{ v \left[ h(l+1)R_v(s) + h(l)\hat{R}_v(s) \right] \frac{\cos \phi}{T} \right\} dsd\phi. \] (21)

For the chip waveform described in (1), the partial autocorrelation functions become \( R_v(s) = s \) and \( \hat{R}_v(s) = T_c - s, \) \( 0 \leq s \leq T_c \) [2]. So, for the BPSK system, the function in (21) becomes

\[ f(v; l, h) = \frac{2}{\pi T_c} \int_{0}^{\pi/2} \int_{0}^{T_c} \cos \left\{ v \left[ h(l+1)s + h(l)(T_c - s) \right] \frac{\cos \phi}{T} \right\} dsd\phi \]

\[ = \frac{2}{\pi T_c} \int_{0}^{\pi/2} \sin \left[ vh(l+1)T_c T^{-1} \cos \phi \right] - \sin \left[ vh(l)T_c T^{-1} \cos \phi \right] \frac{\cos \phi}{v [h(l+1) - h(l)] T^{-1} \cos \phi} d\phi. \]

Applying the trigonometric identity

\[ \sin(2A) - \sin(2B) = 2 \sin(A - B) \cos(A + B), \]

this expression reduces to
\[ f(v; l, h) = \frac{2}{\pi} \int_0^{\pi/2} 2 \sin \left\{ v [h(l + 1) - h(l)] \frac{\cos \phi}{2N} \right\} \cos \left\{ v [h(l + 1) + h(l)] \frac{\cos \phi}{2N} \right\} d\phi \]
\[ = \frac{2}{\pi} \int_0^{\pi/2} \text{sinc} \left\{ v [h(l + 1) - h(l)] \frac{\cos \phi}{2N} \right\} \cos \left\{ v [h(l + 1) + h(l)] \frac{\cos \phi}{2N} \right\} d\phi \quad (22) \]

which is the form obtained in (14) of [2].

For comparative analysis, the Gaussian approximation of the probability of error is derived by applying the average signal-to-noise ratio at the receiver output when there is no power in the faded component of the zeroth path (i.e., \( \gamma_0 = 0 \)). As in [3], SNR\(_0\) is defined by

\[ \text{SNR}_0 = E[Z|b_0 = +1] \{ \text{Var}[Z|b_0 = +1] \}^{-1/2} \]
\[ = \sqrt{E_b} \left\{ E \left[ 1 + \sum_{n=1}^L \gamma_n F_n \right] \right\} \left\{ \text{Var} \left[ \eta_0' + \sqrt{E_b} \left( 1 + \sum_{n=1}^L \gamma_n F_n \right) \right] \right\}^{-1/2} \]
\[ = \sqrt{E_b} \left\{ \text{Var}(\eta_0') + E_b \text{Var} \left[ \sum_{n=1}^L \gamma_n F_n \right] \right\}^{-1/2} \quad (23) \]

Since \( E \left[ \sum_{n=1}^L \gamma_n F_n \right] = 0 \), \( \text{Var} \left[ \sum_{n=1}^L \gamma_n F_n \right] = \sum_{n=1}^L \gamma_n^2 E(F_n^2) \), where

\[ E(F_n^2) = E \left\{ T^{-2} [b_{-1} R_x(\tau_n) + \hat{R}_x(\tau_n)]^2 \cos^2(\phi_n) \right\} \]
\[ = \frac{1}{2} T^{-2} E \left\{ [b_{-1} R_x(\tau_n) + \hat{R}_x(\tau_n)]^2 \right\} \]
\[ = \frac{1}{2} T^{-2} E[R_x^2(\tau_n) + \hat{R}_x^2(\tau_n)] \]
\[ = \frac{1}{2} T^{-2} \frac{1}{M_2 T_c - M_1 T_c} \int_{M_1 T_c}^{M_2 T_c} [R_x^2(\tau) + \hat{R}_x^2(\tau)] d\tau. \]

Making this substitution, (23) becomes

\[ 13 \]
\[ S N R_0 = \sqrt{E_b} \left\{ \frac{N_0}{2} + \frac{E_b}{2T^2 T_c(M_2 - M_1)} \sum_{n=1}^{L} \gamma_n^2 \int_{M_1 T_c}^{M_2 T_c} [R_x^2(\tau) + \dot{R}_x^2(\tau)] d\tau \right\}^{-1/2} \]

\[ = \left\{ \frac{N_0}{2E_b} + \frac{1}{2} \frac{1}{T^2 T_c(M_2 - M_1)} \sum_{n=1}^{L} \gamma_n^2 \int_{M_1 T_c}^{M_2 T_c} [R_x^2(\tau) + \dot{R}_x^2(\tau)] d\tau \right\}^{-1/2} \]

\[ = \left\{ \left( \frac{2E_b}{N_0} \right)^{-1} + \frac{1}{2} [T^2 T_c(M_2 - M_1)]^{-1} \sum_{n=1}^{L} \gamma_n^2 \int_{M_1 T_c}^{M_2 T_c} [R_x^2(\tau) + \dot{R}_x^2(\tau)] d\tau \right\}^{-1/2} \]

(24)

Substituting the expressions for \( R_x \) and \( \dot{R}_x \), the sum within the integral becomes

\[ R_x^2(\tau) + \dot{R}_x^2(\tau) = [C_x(N - l)\dot{R}_\psi(\tau - l T_c) + C_x(N - l - 1)R_\psi(\tau - l T_c)]^2 \]

\[ + [C_x(l)\dot{R}_\psi(\tau - l T_c) + C_x(l + 1)R_\psi(\tau - l T_c)]^2 \]

\[ = \dot{R}_\psi(\tau - l T_c) [C_x^2(N - l) + C_x^2(l)] \]

\[ + 2\dot{R}_\psi(\tau - l T_c) R_\psi(\tau - l T_c) [C_x(N - l)C_x(N - l - 1) + C_x(l)C_x(l + 1)] \]

\[ + R_\psi^2(\tau - l T_c) [C_x^2(N - l - 1) + C_x^2(l + 1)]. \]

Again, letting \( s = \tau - l T_c \) and realizing the expressions for \( R_\psi(s) \) and \( \dot{R}_\psi(s) \),

\[ \int_{M_1 T_c}^{M_2 T_c} [R_x^2(\tau) + \dot{R}_x^2(\tau)] d\tau \]

\[ = \sum_{l=M_1}^{M_2-1} \int_{0}^{T_c} \{(T_c - s)^2[C_x^2(N - l) + C_x^2(l)] + 2(T_c - s)(s)[C_x(N - l)C_x(N - l - 1) + C_x(l)C_x(l + 1)] \}

\[ + C_x(l)C_x(l + 1) + s^2[C_x^2(N - l - 1) + C_x^2(l + 1)] \}

\[ = \sum_{l=M_1}^{M_2-1} \left\{ -[C_x^2(N - l) + C_x^2(l)] \frac{(T_c - s)^3}{3} \right\} \]

\[ + 2[C_x(N - l)C_x(N - l - 1) + C_x(l)C_x(l + 1)] \left( \frac{T_c s^2}{2} - \frac{s^3}{3} \right) \]

14
Thus, the signal-to-noise ratio reduces to

\[
SNR_0 = \left\{ \left( \frac{2\beta}{N_0} \right)^{-1} + \frac{1}{2} \left[ T^2(M_2T - M_1T) \right]^{-1} \sum_{n=1}^{L} \sum_{i=M_1}^{M_2-1} \frac{T^3}{3} \left[ C_x(N - l)C_x(N - l - 1) + C_x(l)C_x(l + 1) + C_x^2(N - l - 1) + C_x^2(l + 1) + C_x^2(l + 1) + C_x^2(l + 1) \right] \right\}^{-1/2}
\]

\[
= \left\{ \left( \frac{2\beta}{N_0} \right)^{-1} + \frac{L\gamma_n^2}{6N} \sum_{i=M_1}^{M_2-1} \left[ C_x(N - l)C_x(N - l - 1) + C_x(l)C_x(l + 1) + C_x^2(N - l - 1) + C_x^2(l + 1) + C_x^2(l + 1) + C_x^2(l + 1) \right] \right\}^{-1/2}.
\]

(25)

When consideration is given to the case \( \gamma_0 \neq 0 \), the Gaussian approximation to the probability of error must be conditioned on the phase shift \( \theta_0 \), as seen in the expression for \( F_0 \). Thus, as in [1], the approximation becomes

\[
P_G = \frac{1}{2\pi} \int_0^{2\pi} Q[SNR_0(1 + \gamma_0 \cos \theta)]d\theta.
\]

(26)
4. Coded Performance Analysis

The performance of the spread-spectrum system is enhanced with the addition of channel coding, as one may anticipate. Due to the hardware constraints for physical application of such systems, it becomes necessary to quantize the continuous receiver output $Z$ in order to examine the effects of coding. We start by defining $W$ as a quantization function of $Z$: that is,

$$W = w_j \quad \text{for} \quad a_j \leq Z < a_{j+1}, \quad j = 1, 2, ..., J$$

where $a_1 = -\infty$ and $a_J = +\infty$. Here, $a_j$ represents the lower quantization threshold for quantization level $j$ and we assume perfect interleaving. For coding with soft decisions there are many levels $j$, while with hard decisions, $j$ is only as high as two. It follows from (13) then that the probability mass function (pmf) for $W$ is given by

$$P\{W = w_j\} = P(a_j \leq Z < a_{j+1}) = P[a_j \leq \sqrt{E}(\eta + b_0 + I) < a_{j+1}]. \quad (27)$$

Here, $E_b$ is replaced by $E$, the energy per code symbol: in general, for rate-$r$ coding, $E = rE_b$. Letting $\phi$ represent the characteristic function for the sum $\eta + I$, and noting that $\phi$ is an even function,

$$P(a_j \leq Z < a_{j+1}|b_0) = \int_{\sqrt{E}/b_0}^{\sqrt{E}/b_0} \frac{1}{\pi} \int_0^{\infty} \phi(u) \cos(gu)du dg. \quad (28)$$

As shown in Appendix A, one may replace $\phi(u) = \phi_n(u)\phi_I(u)$ with $\phi_n(u) - \phi_n(u)[1-\phi_I(u)]$ to avoid subtracting numbers large relative to their difference.
Thus, (28) becomes

$$P(a_j \leq Z < a_{j+1} | b_0) = \frac{1}{\pi} \int_0^\infty u^{-1} d(u) \phi_n(u) du - \frac{1}{\pi} \int_0^\infty u^{-1} d(u) \phi_n(u)[1 - \phi_I(u)] du$$  \hspace{1cm} (29)$$

where $\phi_n$ and $\phi_I$ are given in (16)-(20) and the function $d(u)$ is defined by

$$d(u) = \sin\left(\left(\frac{a_j+1}{\sqrt{E}} - b_0\right) u\right) - \sin\left(\left(\frac{a_j}{\sqrt{E}} - b_0\right) u\right).$$

For channel coding, an upper bound to the probability of error may be determined using $D$-parameter analysis. Denoting the conditional probability mass function by $p(w_j | b_0)$, the distance measure is approximated by

$$D = \sum_{j=1}^J \sqrt{p(w_j | b_0) p(w_j, -1)}$$  \hspace{1cm} (30)$$

for $J$ quantization levels. The probability error bound is then expressed as a polynomial function of $D$. For example, for a channel with rate-$1/2$ convolutional code,

$$P_e \leq 36D^{10} + 211D^{12} + 1404D^{14} + 11633D^{16} + 77433D^{18} + 502690D^{20}$$  \hspace{1cm} (31)$$

while for convolutional coding with rate $1/3$,

$$P_e \leq 11D^{18} + 32D^{20} + 195D^{22} + 564D^{24} + 1473D^{26},$$  \hspace{1cm} (32)$$

where the weight enumerator polynomials are constructed from Tables I and II of [6].
For all of the following examples, the signature sequences were generated from the method for auto-optimal (AO) / least-sidelobe-energy (LSE) maximal-length (m) sequences, as described in [4]. Specifically, the parameters listed in Figure A.1 [4] (or [5] when necessary) were applied to a linear feedback shift register to produce the desired binary chipping code (i.e., +1 and -1 values). These sequences were chosen for their structure simplicity and good correlation properties [4]. As in [1], it is assumed that the relative power in the nth path \( \gamma_n^2 = \gamma^2 \) for \( 1 \leq n \leq L \), but the faded component power in the zeroth path, \( \gamma_0^2 \), does not necessarily equal \( \gamma^2 \). In fact, for many instances \( \gamma_0^2 \) will assume a zero value (i.e., no faded component in the zeroth path). All of the numerical results were evaluated using the computer programs listed in Appendix B.

The first example considers the effect on the average probability of error, determined from the characteristic-function method (15), when the path delay is less than a chipping period. Figure 1 shows a comparison in error probability between two PSK DS/SS systems, each multipath channel with \( L = 5 \) reflected paths, individually having relative power \( \gamma^2 = 0.4 \) and \( \gamma_0^2 = 0.1 \). The signature sequence used is the \( N = 31 \) sequence taken from the first line of Figure A.1 [4]. In the first system, the path delay \( \tau \) is uniformly distributed in \([T_c, T - T_c]\), while in the second, the delay ranges uniformly in \([0, T - T_c]\).
Figure 1: Probability of error for varied path delay uniformity ($N = 31, L = 5, \gamma^2 = 0.4, \gamma_0^2 = 0.1$).
According to the figure, the system performs well when the path delay is at least one chipping period: the probability of error quickly decreases for increasing $E_b/N_0$ values. However, when $\tau \sim \mathcal{U}[0, T - T_z]$, the error probability decreases at a much slower rate and eventually levels off at about $10^{-4}$. This demonstrates the fact that in the second system the multipath interference component arrives too close to the zeroth path signal. As a result, the receiver is unable to reject the delayed path effects.

The second example is a comparison between narrowband and broadband spread-spectrum systems. In the first case (narrowband), a signature sequence of length $N = 127$ is created using shift register polynomial parameter $\mathcal{H} = 301$ and initial register loading $\alpha_0 = 0010010$ [5]. The multipath channel has 5 delayed paths, each with $\tau \sim \mathcal{U}[0, 10T_z]$. For the broadband case, a sequence of length $N = 1023$ is generated using $\mathcal{H} = 2201$ and $\alpha_0 = 0111011010$ [5]. Here, each of the 5 paths is uniformly distributed in $[0, 80T_z]$. This corresponds to the same delay spread if the data rate is as in the $N = 127$ case. In both cases, the faded component of the zeroth path is removed (i.e., $\gamma_0^2 = 0$), and each path has relative power $\gamma^2 = 0.4$. Figure 2 shows the probability of error for each application.

As one might expect, the broadband application outperforms the narrowband one for higher values of $E_b/N_0$. Clearly, as $N$ increases, the chipping period $T_z$ must decrease, assuming constant data pulse length $T$; thus, providing greater multipath rejection capability for the receiver and, hence, lower
Figure 2: Probability of error for narrowband and broadband applications ($L = 5, \gamma^2 = 0.4, \gamma_0^2 = 0$).
error probability. However, such performance comes at the expense of increased bandwidth and system complexity (more shift registers are required for sequence generation).

In the next example, the average probability of error is analyzed for the two systems of Figure 2 as a function of the number of reflected, or delayed, paths $L$. For a fixed $E_b/N_0$ value of 10 dB, Figure 3 demonstrates that this measure of performance increases as more paths are added to the channel.

For the fourth example, the narrowband and broadband spread-spectrum systems are examined for a particular multipath environment. There are a total of $L = 12$ delayed paths, divided into four groups of three paths, each group with a particular relative power for the individual paths, and $\gamma_0^2 = 0$.

For the narrowband ($N = 127$ chips/data bit), there are:

1. 3 paths: $\tau \sim U[0, 4T_c]$, $\gamma^2 = 1$
2. 3 paths: $\tau \sim U[24T_c, 28T_c]$, $\gamma^2 = 0.1$
3. 3 paths: $\tau \sim U[48T_c, 52T_c]$, $\gamma^2 = 0.1$
4. 3 paths: $\tau \sim U[96T_c, 100T_c]$, $\gamma^2 = 0.03$

For the broadband ($N = 1023$ chips/data bit), the length of each data pulse $T$ is maintained from the narrowband case. Thus, the 12 paths for this system are divided as follows:

1. 3 paths: $\tau \sim U[0, 33T_c]$, $\gamma^2 = 1$
2. 3 paths: $\tau \sim U[194T_c, 226T_c]$, $\gamma^2 = 0.1$
3. 3 paths: $\tau \sim U[387T_c, 419T_c]$, $\gamma^2 = 0.1$
4. 3 paths: $\tau \sim U[774T_c, 806T_c]$, $\gamma^2 = 0.03$
Figure 3: Probability of error for increasing number of delayed paths
($E_b/N_0 = 10$ dB, $\gamma^2 = 0.4, \gamma_0^2 = 0$).
Figure 4: Probability of error for a particular multipath environment ($L = 12$ total paths, $\gamma_0^2 = 0$).
As shown by Figure 4, while the probabilities of error for both cases are fairly close at low $E_b/N_0$ values (0-3 dB), their relative difference does increase for higher $E_b/N_0$. Notice, though, that these systems perform rather poorly in comparison to previous situations; $P_e$ lies above $10^{-4}$ even at $E_b/N_0 = 16$ dB. This is due to several factors. First, the path delay region in which a portion of multipath rejection is unattainable (i.e., $\tau < T_c$) also has the greatest amount of relative power for each path ($\gamma^2 = 1$). As a result, this interference component receives the most emphasis of the four. Second, the high number of paths in total (12) should generally imply higher error probability, as proven from Figure 3.

Figure 5 compares three different systems for the length $N = 31$ signature sequence in which path delay is uniformly distributed in $[0, T_c]$, $\gamma_0^2 = 0$, and the sum of the multipath relative powers is constant for each system (i.e., $\sum_{n=1}^{L} \gamma_n^2 = L\gamma^2 = 0.2$). The total number of paths $L$ increases from 2 to 100, as shown. The $L = 100$ curve coincides with the Gaussian approximation to the error probability, expressed in (26), which depends only on $\sum_{n=1}^{L} \gamma_n^2$.

The impact of the multipath interference is demonstrated by the significant difference between the $L = 2$ data and that of $L = 10$ and 100 for $E_b/N_0 > 11$ dB. More importantly, the figure shows that the performance of the characteristic-function method approaches that of the Gaussian model for high values of $L$. In the case of $L = 10$, the characteristic-function model appears Gaussian up to about $E_b/N_0 = 13$ dB, while for $L = 100$, the ab-
Figure 5: Probability of error for characteristic-function method approaching Gaussian approximation in performance $(N = 31, \tau \sim U[0, T_0], \gamma_0^2 = 0)$. 

+ : $L = 2, \gamma^2 = 0.1$  
** : $L = 10, \gamma^2 = 0.02$  
$: L = 100, \gamma^2 = 0.002$  
-- : Gaussian approx.
solute difference in error probability is less than $2 \times 10^{-8}$. From the Central Limit Theorem, one would expect that for sufficiently high $L$ the distribution of the multipath interference is indeed approximated by the Gaussian distribution.

The last two examples examine the performance of DS/SS systems with and without channel coding. In the first, rate-1/2 and 1/3 convolutional coding is applied to a system with $N = 31$ chips/data bit signature sequence and path delay $\tau \sim U[T_c, T - T_c]$; thus, the demodulating sequence has length $N = 62$ or 93 chips/coded symbol, depending upon the chosen rate. Figure 6 compares the probability of error bounds (31,32) for the coded systems with the average error probabilities of uncoded systems with effectively similar chipping sequence lengths. That is, the rate-1/2 system is plotted against an uncoded one having sequence length $N = 63$ chips/data bit ($H = 103$ and $a_0 = 000010$ [4]) and $\tau \sim U[3T_c, 61T_c]$, while the counterpart for the rate-1/3 system has $N = 127$ chips/data bit and $\tau \sim U[5T_c, 123T_c]$, assuming the same data bit duration $T$ of the original system. Both cases of soft and hard decisions are included, using (31)-(32). For soft decisions, a large number (>100) of quantization levels was chosen. The figure also shows the error probability for the system prior to any channel coding as a matter of reference. Comparisons of coded and uncoded systems is somewhat biased because the performance of the coded system is based on an upper bound, while the uncoded system is based on an exact calculation.
Figure 6: Probability of error for DS/SS systems, with and without channel coding ($L = 5, \gamma^2 = 0.4, \gamma_0^2 = 0$).
As expected, the coded systems achieve lower error probabilities for high
$E_b/N_0$ than those of the comparison models. However, at $E_b/N_0$ less than
about 5.5 dB, the rate-1/2 coded system with hard decisions starts to perform
worse. When $E_b/N_0$ is as low as about 1.2 dB, even soft decisions with the
rate-1/3 coded system fail to perform effectively. This is due to the fact that
at sufficiently low input signal-to-noise ratios, channel coding exploits the
weak energy of the data signal and introduces further errors to the system.
As a result, it is not uncommon to even find error probabilities greater than
one (e.g., rate-1/2 coding (hard decision) at $E_b/N_0 = 0$ dB). For the most
part, the systems with soft decisions perform about 3 dB better than hard
decision schemes.

Finally, the channel coding schemes above are applied to the fourth ex-
ample situation in which there is a particular multipath environment for each
system. Figures 7 and 8 show the same type of probability of error data as
in the previous example for the narrowband and broadband spread-spectrum
systems, respectively.

For the narrowband case ($N = 127$), an uncoded system with chipping
sequence length $N = 255$ ($\mathcal{N} = 435$ and $\alpha_0 = 01101111 \ [4]$) is compared with
a rate-1/2 coded system ($N = 254$ chips/information bit). Coding with soft
decisions perform better after $E_b/N_0 = 3.8$ dB, while hard decisions pro-
duce lower error probability only after $E_b/N_0 = 7.3$ dB. The rate-1/3 coded
systems ($N = 381$ chips/information bit) are analyzed against an uncoded
Figure 7: Probability of error for narrowband system and particular multipath environment, with and without channel coding ($L = 12$ total paths, $\gamma_0^2 = 0$).
Figure 8: Probability of error for broadband system and particular multipath environment, with and without channel coding ($L = 12$ total paths, $\gamma_0^2 = 0$).
system with signature sequence $N = 511 (\mathcal{H} = 1041$ and $\alpha_0 = 010101001 [5])$. Here, for $E_b/N_0 > 1.7$ dB, the soft-decision coding outperforms the uncoded system, while hard decisions take effect only after $E_b/N_0 = 4.7$ dB. The threshold distinguishing the effectiveness of soft decisions over hard decisions is at $E_b/N_0 = 2.4$ dB, prior to which the rate-1/3, hard-decision code does better than that with rate-1/2 soft decisions.

For the broadband case ($N = 1023$), the rate-1/2 coded system ($N = 2046$ chips/information bit) is compared with an uncoded system with chipping sequence length $N = 2047 (\mathcal{H} = 5001$ and $\alpha_0 = 110000001000 [5])$. At $E_b/N_0 = 2.8$ dB, soft decisions begin to produce lower error probability, while the same is true for hard decisions at $E_b/N_0 > 5.5$ dB. The rate-1/3 coded system has bandwidth comparable to an uncoded system with $N = 4095 (\mathcal{H} = 14501$ and $\alpha_0 = 10111100001)$. In this case, for $E_b/N_0 > 1.2$ dB, the coded system with soft decisions performs better than the uncoded comparison, while hard decisions do not take effect until after $E_b/N_0 = 3.9$ dB. For this particular system, the rate-1/3 code with hard decisions is more effective than the rate-1/2 code with soft decisions for $E_b/N_0 < 1.1$ dB.
6. Conclusions

From Figure 1, it is clear that the DS/SS system is only effective at rejecting multipath when the path delays are at least one chip period. One method of reducing the average probability of error is to increase the processing gain $N$. However, this comes at the expense of bandwidth and transmitter complexity (i.e., more shift registers required for signature sequence generation). When more delayed paths are added to the system, the multipath interference increases and system performance degrades rapidly. For a sufficiently high number of paths, the performance obtained via the characteristic-function method approaches that of the Gaussian approximation, as expected from the Central Limit Theorem.

The DS/SS system is significantly enhanced with the application of channel coding. In fact, according to Figure 6, the probability of error drops to $10^{-10}$ at about $E_b/N_0 = 6.3 \text{ dB}$ for a system with rate-1/3 convolutional coding and soft decisions applied. For the spread-spectrum systems with a particular multipath environment, there is a significant difference between the narrowband and broadband cases at low error probability (high $E_b/N_0$). However, for relatively higher probability of error (low $E_b/N_0$), the broadband system provides a 1-2 dB improvement over the narrowband counterpart.
7. References


Appendix A

Characteristic-Function Approach To Probability of Error

In this appendix, for purposes of completeness, we derive the expression in [1] for the average probability of error, given in terms of the characteristic functions of the multipath interference and noise. From (14) of the text, the error probability is expressed as

\[ P_e = \frac{1}{2} - \frac{1}{2} P(-1 < \eta + I) \leq 1). \]  \hspace{1cm} (33)

Letting \( G = \eta + I \), and denoting the probability density of \( G \) by \( f_G(g) \), (33) becomes

\[ P_e = \frac{1}{2} - \frac{1}{2} \int_{-1}^{1} f_G(g) dg. \]  \hspace{1cm} (34)

Since the random variables \( \eta \) and \( I \) are mutually independent and have symmetric distributions, their characteristic functions, as well as that of their sum, must be real and even. Denoting the characteristic functions of \( \eta \), \( I \), and \( \eta + I \) by \( \phi_\eta \), \( \phi_I \), and \( \phi \), respectively, \( f_G(g) \) is given by

\[ f_G(g) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-j\xi g} \phi(u) du \]
\[ = \frac{1}{\pi} \int_{0}^{\infty} \phi(u) \cos(\xi u) du \]

since \( f_G(g) \) must also be real-valued. Substituting into (34),

\[ P_e = \frac{1}{2} - \frac{1}{2} \int_{-1}^{1} \frac{1}{\pi} \int_{0}^{\infty} \phi(u) \cos(\xi u) du \] \hspace{1cm} \int dg \]
To avoid subtracting very large numbers relative to their difference, one may replace \( \phi(u) = \phi_n \phi_I \) with \( \phi_n - \phi_n[1 - \phi_I] \). Thus,

\[
\tilde{P}_e = \frac{1}{2} - \frac{1}{\pi} \int_0^\infty u^{-1} \sin(u) \phi_n(u) du \\
+ \frac{1}{\pi} \int_0^\infty u^{-1} \sin(u) \phi_n(u)[1 - \phi_I(u)] du.
\]

(35)

In the absence of interference (i.e., \( I = 0 \)), \( \phi_I = 1 \) and error is due only to the presence of Gaussian noise. Denoting this noise error probability by \( P_n \), (34) becomes

\[
\tilde{P}_e = P_n + \pi^{-1} \int_0^\infty u^{-1} \sin(u) \phi_n(u)[1 - \phi_I(u)] du.
\]

(36)

Since \( \eta \) is Gaussian, \( P_n \) is given by \( Q(\sigma^{-1}) \), where \( \sigma^2 \) is defined as the variance of \( \eta \). The result is (15) of the text.
Appendix B

Computer Programs For Characteristic-Function Method

The following programs provide simulations for the average probability of error (15), the Gaussian approximation (26), and error bounds for coding (31,32). The computer code is written in MATLAB version 4.0 and all integrations are approximated applying Simpson’s method.

The first program, PROBERROR.m, computes the average probability of error for both the characteristic-function method and the Gaussian approximation. The user is able to select the desired signature sequence from a menu; the program assumes existing data files containing the sequence itself and all the appropriate autocorrelation functions. The user may also create one or more multipath environments, each with a desired range for the path delay. The following function files are called within the main program:

SIMP.m – A Simpson’s approximation to the $\phi_0$ integral.

PHI.m – Computes $\tilde{\phi}$; calls function file FF.m which provides Simpson’s method for function $f(v;l,h)$.

The second program, PROBMASS.m, first determines the probability mass function for the quantized receiver output (29). This is then used to compute the distance parameter $D$ (30), which is applied in approximating upper probability bounds for channel coding purposes; the program considers
convolutional coding with rates 1/2 and 1/3. The pmf data is calculated from the same two function files of the previous program.
clear; format long e;
k = menu('Choose sequence length', 'N=31', 'N=63', 'N=127', 'N=255', 'N=511', ...
    'N=1023', 'N=2047', 'N=4095');
if k == 1,
    load code_31 % each signature sequence file contains the actual
    % sequence (code) and all autocorrelation functions:
elseif k == 2,
    load code_63 % aperiodic: R
elseif k == 3,
    load code_127 % odd: thetahat
elseif k == 4,
    load code_255
elseif k == 5,
    load code_511
elseif k == 6,
    load code_1023
elseif k == 7,
    load code_2047
else
    load code_4095
end
disp('Enter the requested data:')
gamma0sq = input('gamma_0-squared = '); % assumed to be constant for all
    % multipath environments.
gamma0 = sqrt(gamma0sq);
sit = input('How many different delay situations are there? ');
for j = 1:sit
    disp('situation #'); disp(j);
    L(j) = input('no. of paths L = ');
    gammasq(j) = input('gamma-squared = ');
    M1(j) = input('constant multiple M1 = ');
    M2(j) = input('constant multiple M2 = ');
    gamma(j) = sqrt(gammasq(j));
end;
N = length(code); % length of sequence
data = input('How many Eb/No data values? ');
ssuml = zeros(1, data); ssun2 = zeros(1, data);
for d = 1:data
    EbN0 dB(d) = input('Enter Eb/N0 value (dB): ');
    b = input('upper-limit of integration = ');
    % computation code
end

n = input('no. of pts. (must be even) for integration = ');
h = b/n; EbN0(d) = 10^(EbN0_db(d)/10);
sigmasq(d) = 1/(2*EbN0(d)); sigmainv(d) = 1/sqrt(sigmasq(d));

Q(d) = Qfunct(sigmainv(d)); % compute the Q-function of sigma-
% inverse.

i = [0:n]; % Simpson's method starts here, done using vector form.
u = 1E-100 + i*h;
phi_eta = exp(-(u.'^2) * (sigmasq(d))/2); % SIMP.m is a function
phi_tilde_0 = 2/pi * simp(gamma0*u, pi/2, 20); % Simpson's method for
% phi_tilde_0.

phi_tilde = phi(gamma' * u, L, M1, M2, N, theta, thetahat); % see separate
% listing for
% function PHI.m.

phi_I = phi_tilde_0 .* phi_tilde;
g = ones(1, n+1) ./ u .* sin(u) .* (1-phi_I) .* phi_eta;

coeff = 2 * (rem(i,2)+1); % coefficients for Simpson's summation.
coeff(1) = 1; coeff(n+1) = 1;
ssum(d) = sum(coeff .* g);
simpson(d) = h/3 * ssum(d);
proberr(d) = Q(d) + 1/pi * simpson(d); % this is eq. (15) of text.

% This section computes the Gaussian approximation
sum2 = 0;
for j = 1:sit
    sum1 = 0;
    for lg = M1(j):M2(j)-1
        sum1 = sum1 + R(2*N+1-lg)*R(2*N-lg) + R(N+1+lg)*R(N+2+lg)...
            + R(2*N+1-lg)^2 + R(2*N-lg)^2 + R(N+1+lg)^2 + ... 
            R(N+2+lg)^2;
    end
    sum2 = sum2 + 1/6 * L(j)*sigmasq(j)/(N^2*(M2(j)-M1(j))) * sum1;
end
SNR_0 = 1 / sqrt(sigmasq(d) + sum2);

ng = 500; % choose 500 points for Simpson's method integration.
ig = [0:ng]; hg = pi/ng; % may use pi as an upper limit since function
% for integration is even

thetag = ig*hg;
g = Qfunct(SNR_0*(1 + gamma0*cos(thetag)));
coefg = 2 * (rem(ig,2)+1);
coefg(1) = 1; coefg(ng+1) = 1;
ssumg(d) = sum(coefg .* g);
simpsong(d) = hg/3 * ssumg(d);
proberrg(d) = 1/pi * simpsong(d); % 1/pi factor instead of 1/(2pi)
% since integrating only to pi; this
% is eq. (26).

disp('ave. prob. of error for char-funct method = '); disp(proberr(d))
disp('Gaussian approximation to prob. of error = '); disp(proberrg(d))
end
function simpson = simp(const,b,n)

% SIMP Simpson's rule for phi_tilde_0 characteristic function.
% Input: gamma_0*u (const), upper-limit (b) and number of points for
% integration (n).
% Output: Simpson's approximation to the integral.

h = b/n; ssum = 0;
i = [0:n];
theta = i*h;
g = cos(const' * cos(theta)); % function for integration.

coeff = 2 * (rem(i,2)+1); % coefficients for approximation.
coeff(1) = 1; coeff(n+1) = 1;
ssum = coeff * g';
simpson = h/3 * ssum; % Simpson's approximation
function out = phi(var,L,M1,M2,N,theta,thetahat)

% PHI Characteristic function phi_tilde(u). Function f for theta and thetahat
% is called by FF.m.
% % Input: variable (var), no. of reflected paths (L), constant multiples of
% % Tc (M1,M2), length of signature sequence (N), and autocorrelation
% % vectors theta and thetahat.
% % Output: the value of the characteristic function.

s = length(L); % number of multipath environments.

prod1 = 1; prod2 = 1;
for j = 1:s
    ssum1 = 0; ssum2 = 0;
    for l = M1(j):M2(j)-1
        f = ff(var(j,:),l,theta,N); % compute f for theta and
        % theta hat
        f_hat = ff(var(j,:),l,thetahat,N);
        ssum1 = ssum1 + f;
        ssum2 = ssum2 + f_hat;
    end
    out1 = (1/(M2(j)-M1(j)) * ssum1).^L(j);
    out2 = (1/(M2(j)-M1(j)) * ssum2).^L(j);
    prod1 = prod1 .* out1;
    prod2 = prod2 .* out2;
end

out = 0.5 * (prod1 + prod2); % this corresponds to phi_tilde(u).
function f = ff(v,k,t,N)

% FF Function f for vectors theta and thetahat. Considers zeroeth index for
% autocorrelation vectors. Input is real variable (v), index (k),
% autocorrelation vector (t), and length of code sequence (N). Output is
% Simpson's formula approximation.

b = pi/2; n = 20;    % only 20 points are necessary for good approximation.
% h = b/n;
i = [0:n]; l = k+1;
theta = i*h;
g1 = sinc(v'*(t(l+1)-t(l))*cos(theta)/(2*pi*N)); % SINC.m simply performs the
% sinc function for given
% variable.

g2 = cos(v'*(t(l+1)+t(l))*cos(theta)/(2*N));
g = g1 .* g2;
coeff = 2 * (rem(i,2)+1);
coeff(1) = 1; coeff(n+1) = 1;
ssum = coeff * g';
simpson = h/3 * ssum; f = 2/pi * simpson;        % this is the function f.
A program to compute the probability mass function of $Z$, $P(a < Z < b | b_0)$, of a multipath DS/SS system with BPSK modulation. Path delay is uniform from one multiple of $T_c$ to another, for a given number of multipath environments. User selects desired chipping sequence from a menu. Here, gamma0-squared($\gamma_0^2$) is assumed to be zero, but may be reset to constant for all delay situations. The pmf is used, in turn, to determine the Hamming D-parameter and apply toward computing the probability error bounds for specific channel coding.

**Input**: for each delay environment -- number of paths ($L$), gamma-squared ($\gamma_0^2$), $T_c$ multiples ($M_1, M_2$); total number of ranges for receiver output $Z$, epsilon increment, $E_b/N_0$, upper limit and no. of points for integration using Simpson's method.

**Output**: probability error bounds for rate-$1/2$ and $1/3$ convolutional codes.

```matlab
% PROBMASS.m
% A program to compute the probability mass function of Z, P(a<Z<b|b0), of a
% multipath DS/SS system with BPSK modulation. Path delay is uniform from one
% multiple of Tc to another, for a given number of multipath environments. User
% selects desired chipping sequence from a menu. Here, gamma0-squared(gamma0sq)
% is assumed to be zero, but may be reset to constant for all delay situations.
% The pmf is used, in turn, to determine the Hamming D-parameter and apply
% toward computing the probability error bounds for specific channel coding.

% Input: for each delay environment -- number of paths (L), gamma-squared
% (gamma0sq), Tc multiples (M1,M2);
% total number of ranges for receiver output Z, epsilon increment, Eb/No,
% upper limit and no. of points for integration using Simpson's method.

% Output: probability error bounds for rate-1/2 and 1/3 convolutional codes

clear; format long e;

k = menu('Choose sequence length', 'N=31', 'N=127', 'N=1023');
if k == 1,
    load code_31
elseif k == 2,
    load code_127
else
    load code_1023
end
disp('Enter the requested data:"

sit = input('How many different delay situations are there? ');
for j = 1:sit
    disp('situation #'); disp(j);
    L(j) = input('no. of paths L = ');
    gamma0sq(j) = input('gamma-squared = ');
    disp('Path delay tau - U[M1*Tc, M2*Tc]:')
    M1(j) = input('constant multiple M1 = ');
    M2(j) = input('constant multiple M2 = ');
    gamma0sq = 0; % assume gamma0-squared is zero throughout.
    gamma(j) = sqrt(gamma0sq(j)); gamma0 = sqrt(gamma0sq);
end;
N = length(code);
num = input('How many Eb/No values?');
for g = 1:num
    EbNO_dB(g) = input('Enter Eb/No (dB):');
    eps = input('Enter increment epsilon:'); % increment used in mapping
    data = input('How many total ranges for Z (even)? ');
    p = [-data/2:eps/10]
    Eb(g) = 10^(EbNO_dB(g)/10); N0(g) = 1; % the energy per information bit
end
```

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\[ E_2(g) = \frac{E_b(g)}{2}; \quad E_3(g) = \frac{E_b(g)}{3}; \]
\[ \text{this is the energy per coded symbol for rates-1/2 and 1/3 codes.} \]

\[ \text{ENO}_2(g) = \frac{(E_b(g)/N_0(g))/2; \quad \text{ENO}_3(g) = \frac{(E_b(g)/N_0(g))/3;} \]

\[ b_b = \text{input('upper-limit of integration =')}; \]
\[ n = \text{input('no. of pts. (must be even) for integration =')}; \]
\[ h = b_b/n; \]
\[ \text{sigmasq}_2(g) = \frac{1}{2*\text{ENO}_2(g)}; \quad \text{sigmasq}_3(g) = \frac{1}{2*\text{ENO}_3(g)}; \]
\[ \text{sigmainv}_2(g) = \frac{1}{\sqrt{\text{sigmasq}_2(g)}}; \quad \text{sigmainv}_3(g) = \frac{1}{\sqrt{\text{sigmasq}_3(g)}}; \]
\[ \text{d2} = \text{eps} \times p \times \sqrt{E_2(g)}; \quad \text{d3} = \text{eps} \times p \times \sqrt{E_3(g)}; \]
\[ \text{a2 = d2(1:}\text{data}; \quad \text{a3 = d3(1:}\text{data}; \]
\[ \text{b2 = d2(2:}\text{data}+1); \quad \text{b3 = d3(2:}\text{data}+1); \]
\[ \text{a and b the range limits for the Z values. Throughout, a 2-suffix refers to a variable for rate-1/2 code, while a 3-suffix refers to that for rate-1/3 code.} \]

\[ i = [0:n]; \]
\[ u = 1E-100 + i*h; \]
\[ \text{uinv = ones(1,n+1)}/u; \quad \text{uinvd = uinv}; \]
\[ \text{phi}_\text{eta}_2 = \exp(-(u.*2) \times \text{sigmasq}_2(g)/2); \]
\[ \text{phi}_\text{etad}_2 = \text{phi}_\text{eta}_2; \quad \text{phi}_\text{etad}_3 = \text{phi}_\text{eta}_3; \]
\[ \text{phi}_\text{tilde}_0 = 2\pi \times \text{simp(gamma0*u}, \pi/2, 20); \]
\[ \text{phi}_\text{tilde} = \text{phi(gamma}^*u,L,M1,M2,N,\theta,\theta_{\text{that}}); \]
\[ \text{phi}_I = \text{phi}_\text{tilde}_0 \times \text{phi}_\text{tilde}; \quad \text{phi}_I = \text{phi}_I; \]
\[ \text{for m = 1:}\text{data}-1 \]
\[ \text{uinvd = [uinvd; uinv]}; \]
\[ \text{phi}_\text{etad}_2 = [\text{phi}_\text{etad}_2; \text{phi}_\text{eta}_2]; \]
\[ \text{phi}_\text{etad}_3 = [\text{phi}_\text{etad}_3; \text{phi}_\text{eta}_3]; \]
\[ \text{phi}_I = [\text{phi}_I; \text{phi}_I]; \]
\[ \text{end} \]

\[ \text{coeff} = 2 \times (\text{rem(i,2)+1}); \]
\[ \text{coeff(1) = 1; coeff(n+1) = 1}; \]
\[ \text{ssum22pos} = \text{coeff} \times \text{g22pos';} \quad \text{ssum13pos} = \text{coeff} \times \text{g13pos'}; \]
\[ \text{ssum22pos} = \text{coeff} \times \text{g22pos';} \quad \text{ssum23pos} = \text{coeff} \times \text{g23pos'}; \]
ssum12neg = coeff * g12neg'; ssum13neg = coeff * g13neg';
ssum22neg = coeff * g22neg'; ssum23neg = coeff * g23neg';
simpson12pos = h/3 * ssum12pos; simpson13pos = h/3 * ssum13pos;
simpson22pos = h/3 * ssum22pos; simpson23pos = h/3 * ssum23pos;
simpson12neg = h/3 * ssum12neg; simpson13neg = h/3 * ssum13neg;
simpson22neg = h/3 * ssum22neg; simpson23neg = h/3 * ssum23neg;
probmass2pos = 1/pi*simpson12pos - 1/pi*simpson22pos;
probmass3pos = 1/pi*simpson13pos - 1/pi*simpson23pos;

% the probability mass function is assumed to be zero for the final
% value of Z and thereafter; probmass... is eq. (29) for a and b values
% above.

probmass2pos(data+1) = 0; probmass3pos(data+1) = 0;
probmass2neg = 1/pi*simpson12neg - 1/pi*simpson22neg;
probmass3neg = 1/pi*simpson13neg - 1/pi*simpson23neg;
probmass2neg(data+1) = 0; probmass3neg(data+1) = 0;

% this shows how close the sum of the pmfs is to 1.

disp('summation error (1/2) pos = '); disp(1-sum(probmass2pos));
disp('summation error (1/3) pos = '); disp(1-sum(probmass3pos));
disp('summation error (1/2) neg = '); disp(1-sum(probmass2neg));
disp('summation error (1/3) neg = '); disp(1-sum(probmass3neg));

stairs(d2,probmass2pos); pause % plot the pmf in stairs mode.
stairs(d2,probmass2neg); pause
stairs(d3,probmass3pos); pause
stairs(d3,probmass3neg);
D2(g) = sum(sqrt(probmass2pos .* probmass2neg)); % this is the
% D-parameter
D3(g) = sum(sqrt(probmass3pos .* probmass3neg));

% the following formulas are approximations to the probability error
% bounds for rate-1/2 and rate-1/3 convolutional codes -- eqs. (31,32).

bound2(g) = 36*D2(g)^(10) + 211*D2(g)^(12) + 1404*D2(g)^(14)
            + 11633*D2(g)^(16) + 77433*D2(g)^(18) + 502690*D2(g)^(20);

bound3(g) = 11*D3(g)^(18) + 32*D3(g)^(20) + 195*D3(g)^(22) + 564*D3(g)^(24)
            + 1473*D3(g)^(26);

disp('error bound for rate-1/2 code = '); disp(bound2(g));
disp('error bound for rate-1/3 code = '); disp(bound3(g));