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Packet Error Rate Analysis of Uncoded Schemes in Block-Fading Channels using Extreme Value Theory

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Abstract—We present a generic approximation of the packet error rate (PER) function of uncoded schemes in the AWGN channel using extreme value theory (EVT). The PER function can assume both the exponential and the Gaussian $Q$-function bit error rate (BER) forms. The EVT approach leads us to a best closed-form approximation, in terms of accuracy and computational efficiency, of the average PER in block-fading channels. The numerical analysis shows that the approximation holds tight for any value of SNR and packet length whereas the earlier studies approximate the average PER only at asymptotic SNRs and packet lengths.

Index Terms—Packet error rate, block-fading channels, extreme value theory

I. INTRODUCTION

In packet radio systems, the packet error rate (PER) analysis has practical significance for reliability and throughput estimation. Other than noise and packet collisions, therein, modeling the packet errors due to present fading conditions is an important problem. The time scale of fading relative to the bit or packet duration influences the selection of a packet error model. If the wireless channel remains constant across the length of a packet and the consecutive packets observe independent channel realizations, the packet error model must assume constant (block) fading for all the bits in a packet.

Block-fading characterizes the wireless channels that experience slowly varying fading conditions. The PER evaluation in block-fading channels is studied extensively. However, for any modulation and coding scheme, its exact evaluation is complex and usually loose bounding methods as Jensen’s inequality and Chernoff upper bound are employed [1].

In [2], the average PER in Rayleigh block-fading, for both uncoded (commonly used in short-range radio systems) and coded schemes, is tightly upper bounded by $1 - \exp(-\omega_0/\gamma)$, where $\gamma$ is average signal to noise ratio (SNR) and $\omega_0$ corresponds to the inverse coding gain. By definition, $\omega_0 = \int_0^\infty P_e(\gamma) d\gamma$, where $P_e(\gamma)$ is the PER in the AWGN channel. A log-domain linear approximation of $\omega_0$, for uncoded FSK, is proposed in [1]. In [3], the same model is utilized to estimate the parameters of $\omega_0$ for different modulation schemes. However, the upper bound in [2] and the approximations in [1] [3] are tight only in asymptotic regime: that is, when the average SNR is high or the packet length is large.

II. SYSTEM MODEL

Let $x$ be the bits in a packet transmitted over a block-fading channel and $y$ be the received symbols. Then, the input $x$ and the output $y$ of the channel are related as,

$$y = hx + n \quad (1)$$

where $h$ is the instantaneous fading coefficient. It is a zero-mean circularly symmetric complex Gaussian (CSCG) random variable with unit-variance i.e., $E[|h|^2] = 1$. While, $n$ is a sequence of mutually independent, zero-mean, CSCG noise with power $N_0$. If $P$ is the power per transmit bit, $\gamma = |h|^2 P/N_0$ is the instantaneous received SNR and the average SNR is,

$$\bar{\gamma} = E[\gamma] = E[|h|^2 P/N_0] = P/N_0 \quad (2)$$

The average PER over a block-fading channel, $\hat{P}_e(\gamma)$, is computed by integrating the PER in the AWGN, denoted as $\hat{P}_e(\gamma)$, over the distribution of received SNR, $p(\gamma; \bar{\gamma})$ [2],

$$\hat{P}_e(\gamma) = \int_0^\infty \hat{P}_e(\gamma)p(\gamma; \bar{\gamma})d\gamma. \quad (3)$$

Depending on the radio propagation environment, the fading magnitude $|h|$ and consequently $\gamma$ will take different distributions. Whereas, for an $N$-bit packet, $\hat{P}_e(\gamma)$ is defined as,

$$\hat{P}_e(\gamma) = 1 - \left(1 - b_e(\gamma)\right)^N \quad (4)$$

where $b_e(\gamma)$ is the modulation-dependent instantaneous BER. We consider $b_e(\gamma)$ with the following generic forms,

$$b_e(\gamma) = c_m Q\left(\sqrt{k_m \gamma}\right) \quad (5)$$

$$b_e(\gamma) = c_m \exp\left(-k_m \gamma\right) \quad (6)$$

where $c_m$ and $k_m$ are the modulation-specific constants. The modulation schemes such as M-ASK, M-PAM, MSK, M-PSK and M-QAM have the BER form of (5) whereas $Q(\cdot)$ is the Gaussian $Q$-function. Whereas, FSK and DPSK non-coherent modulations are described by the BER form in (6) [2].
III. EVT BASED PER FUNCTION APPROXIMATION

The evaluation of $\hat{P}_c(\gamma)$ is difficult because of $\hat{P}_c(\gamma)$ expression involving a polynomial of degree $N$. In what follows, we provide a generic asymptotic approximation of (4) using extreme value theory (EVT) for the BER expressions described in (5) and (6) that allows to evaluate $\hat{P}_c(\gamma)$ easily.

**Proposition 1:** The PER in the AWGN channel for the BER functions in (5) and (6), as the packet length $N \to \infty$, is asymptotically approximated by the Gumbel distribution for the minimum, i.e.,

$$\hat{P}_c(\gamma) \approx 1 - \exp \left( - \exp \left( - \frac{\gamma - a_N}{b_N} \right) \right)$$

(7)

where $a_N$ and $b_N > 0$ are the normalizing constants.

**Proof:** For a packet of length $N$, the PER in (4) for both (5) and (6) can be written as,

$$\hat{P}_c(\gamma) = 1 - \left( 1 + c_m (F_X(x) - 1) \right)^N$$

(8)

where $F_X(x)$ is the cumulative distribution function (CDF) of the standard normal distribution with $x = \sqrt{k_m \gamma}$ for the BER in (5) while it is the CDF of the exponential distribution with $x = \gamma$ and $\lambda = k_m$ for (6).

For $c_m = 1$ (i.e., BPSK modulation), (8) can be expressed in terms of $[F_X(x)]^N$ which can be approximated, as discussed later, with an extreme value distribution. In order to extend the same approach to any $c_m$, we manipulate (8) as,

$$\hat{P}_c(\gamma) \approx 1 - \left( 1 + c_m \log F_X(x) \right)^N$$

$$\approx 1 - \left( c_m \log F_X(x) \right)^N$$

$$\approx 1 - \left( F_X(x) \right)^{Nc_m}$$

(9)

where $\log(\cdot)$ is the natural logarithm. Here, the first approximation results from the inequality $\log y \leq y - 1$ for $y > 0$, the second from $e^{y} \geq 1 + y$ and the last approximation uses the identity $y^{a} = e^{a \log y}$. The approximation in (9) is quite tight and its accuracy increases as $N$ and $\gamma \to \infty$.

By letting $N' = Nc_m$, we handle $[F_X(x)]^{N'}$ in (9) separately. Note that,

$$[F_X(x)]^{N'} = \prod_{i=1}^{N'} \Pr(X_i \leq x) = \Pr \left( \max_{1 \leq i \leq N'} \{ X_i \} \leq x \right)$$

(10)

implies that $[F_X(x)]^{N'}$ can be approximated by one of the extreme value distributions: Gumbel, Fréchet, Weibull [4]. To assimilate this, assume $X_1, \cdots, X_N$ the i.i.d. random variables drawn from a common distribution function $F(x)$.

Let $M_N = \max X_i$ denotes the maximum of first $N$ random variables. If there exist constants $a_N$ and $b_N > 0$, and a non-degenerate limit distribution $G(x)$ such that the CDF of normalized $M_n$ converges to $G(x)$,

$$\lim_{N \to \infty} \Pr \left( \frac{M_N - a_N}{b_N} \leq x \right) \to G(x)$$

(11)

then $F(x)$ is said to be in the domain of attraction of $G(x)$.

In order to find the exact limiting distribution, we recall the sufficient condition, stated in [4] [5], for a distribution function $F(x)$ belonging to the domain of attraction of the Gumbel distribution. Let $\omega(F) = \sup \{ x : F(x) < 1 \}$, and assume that there is a real number $x_1$ such that for all $x_1 \leq x < \omega(F)$, $f(x) = F'(x)$ and $F''(x)$ exist, and $f(x) \neq 0$. If

$$\lim_{x \to \omega(F)} \frac{d}{dx} \left[ \frac{1 - F(x)}{f(x)} \right] = 0$$

(12)

then $F(x)$ converges to the Gumbel distribution as $N \to \infty$. The Gumbel distribution function is $G(x) = \exp(-\exp(-\frac{x - \theta}{\theta}))$ and constants $a_N$ and $b_N$ are,

$$a_N = F^{-1}(\frac{1}{N})$$

$$b_N = F^{-1}(\frac{1}{Ne}) - a_N$$

(13)

where $e$ is the base of the natural logarithm, and $F^{-1}(\cdot)$ denotes the inverse of $F(x)$.

It is straightforward to show that the normal and the exponential distributions satisfy the sufficient condition stated in (12). By replacing $[F_X(x)]^{N'}$ in (9) with the Gumbel distribution function completes the proof.

The constants $a_N$ and $b_N$ for the normal distribution can be determined from (13) by using $N = N c_m$ and the quantile function of the normal distribution, $F^{-1}(p) = \sqrt{2 \pi} \text{erf}^{-1}(2p - 1)$. As $x = \sqrt{k_m \gamma}$, the transformation for $\gamma$ gives the constants,

$$a_N = \frac{2}{k_m} \left( \text{erf}^{-1}(1 - \frac{2}{N c_m}) \right)^2$$

$$b_N = \frac{2}{k_m} \left( \text{erf}^{-1}(1 - \frac{2}{N c_m}) \right)^2 - a_N$$

(14)

Similarly for the exponential case, from (13) and the quantile function of the exponential distribution, $F^{-1}(p; k_m) = -\log(1-p)/k_m$, we have,

$$a_N = \frac{\log(N c_m)}{k_m}, b_N = \frac{1}{k_m}$$

(15)

In next section, we show that the approximation in (7) can easily be evaluated under a general fading distribution.

IV. PER OVER BLOCK-FADING CHANNELS

We consider Nakagami-$m$ fading model for which the SNR, $\gamma$, is gamma distributed with the probability density function,

$$p(\gamma; \gamma) = \frac{m^m e^{-m \gamma}}{\Gamma(m)} \gamma^{m-1}$$

(16)

where $0.5 \leq m < \infty$ is the fading parameter and $\Gamma(\cdot)$ is the standard gamma function [6, p.892]. Using (7) and (16) in (3),

$$\hat{P}_c(\gamma) \approx 1 - \frac{m^m}{\Gamma(m)} \int_0^\infty e^{-e^{-\frac{\gamma-a_N}{b_N}}-\frac{m}{m}} d\gamma$$

(17)

Let $s = \frac{m}{m}$ and $G(\gamma) = e^{-e^{-\frac{\gamma-a_N}{b_N}}}$, then the integral in (17) is recognized as the Laplace transform of $\gamma^{-1} \Gamma(\gamma)$, i.e.,

$$\hat{P}_c(\gamma) \approx 1 - \frac{m^m}{\Gamma(m)} \int_0^\infty e^{-s\gamma} \gamma^{-1} \Gamma(\gamma) d\gamma$$

(18)

Using the Laplace transform of the PDF of the Gumbel distribution, $\mathcal{L}\left\{g(\gamma)\right\} = e^{-a_N/s} \Gamma(1 + b_N s)$, and the Laplace
transform properties, \( \mathcal{L}(G(\gamma)) = \frac{\mathcal{L}(g(\gamma))}{s} \) and \( \mathcal{L}\{t^n f(t)\} = (-1)^n \frac{d^n}{ds^n} \mathcal{L}\{f(t)\} \), (18) is evaluated as,

\[
P_e(\gamma) \approx 1 - m^m (-1)^{m-1} \frac{d^{m-1}}{ds^{m-1}} \left( e^{-sN} \Gamma(1 + sb_N) \right)
\]

(19)

For \( m = 1 \) (Rayleigh fading), (19) easily reduces to,

\[
P_e(\gamma) \approx 1 - e^{-\frac{sN}{\gamma}} \left( 1 + \frac{b_N}{\gamma} \right).
\]

(20)

By evaluating the derivative in (19) for \( m = 2 \) and \( m = 3 \),

\[
P_e(\gamma) \approx 1 - e^{-\frac{sN}{\gamma}} \left( 1 + \frac{2b_N}{\gamma} \right) \psi \left( 1 + \frac{2b_N}{\gamma} \right)
\]

(21)

\[
P_e(\gamma) \approx 1 - \frac{e^{-3sN/\gamma}}{\gamma^3} \left( 1 + \frac{3b_N}{\gamma} \right)^2 \left( 1 + \frac{2a_N}{\gamma} - \frac{2b_N}{\gamma} \right) \psi \left( 1 + \frac{2b_N}{\gamma} \right) + \frac{9b_N^2}{\gamma^2} \psi \left( 1 + \frac{3b_N}{\gamma} \right) + \frac{9b_N^2}{\gamma^2} \psi \left( 1 + \frac{3b_N}{\gamma} \right)^2 \psi \left( 1 + \frac{3b_N}{\gamma} \right)
\]

(22)

where \( \psi(\cdot) \) and \( \psi(1, \cdot) \) are the digamma and trigamma functions respectively.

V. NUMERICAL RESULTS AND DISCUSSION

We validate the proposed approximation of the average PER for uncoded FSK, BPSK, and 16-QAM modulation schemes in block-fading. For this purpose, the approximated average PER in (19) is validated against the numerical evaluation of (3), and the approximation error is compared with earlier studies.

The BER functions for non-coherent FSK and BPSK schemes are \( \frac{1}{2} \exp(-\gamma/2) \) and \( Q(\sqrt{2\gamma}) \) respectively. The approximated average PER for these schemes is shown in Fig. 1 for \( m = 1 \), and depicts perfect matching to the numerical results for small and large packets. For 16-QAM, we use the BER approximation for M-QAM, \( \frac{2}{k} \left( 1 - \frac{1}{M} \right) Q(\sqrt{2\gamma M/k}) \), where \( M \) is the constellation size and \( k = \log_2 M \) is the number of bits per symbol [7]. The average PER for 16-QAM using (19) shows the same accuracy over \( m, N \) and \( \gamma \) in Fig. 2.

It is worth noting that one can easily find an exact expression of the average PER for FSK by using the binomial expansion of (4) with the BER in (3). For instance, in Rayleigh fading with \( p(\gamma; \tilde{\gamma}) = \exp(-\gamma/\tilde{\gamma}) \), we have,

\[
P_e(\gamma) = 1 - \sum_{n=0}^{N} \frac{N!}{n!(N-n)!} (-1)^n \left( \frac{c_m}{1 + nk_m} \right)^n.
\]

(23)

However, this expression requires the addition of terms, with alternating sign, each obtained from the multiplication of a very large number with a very small number. Consequently, it is numerically difficult to evaluate (23) for large values of \( N \).

Similarly, the Gaussian \( Q \)-function approximations (e.g., [8], [9]) experience the same above-stated difficulty in evaluating the average PER of modulation schemes described by (5). The other approach is to utilize the approximations for an integer power of the Gaussian \( Q \)-function (e.g., see [10]). Such approximations are developed to evaluate the average symbol error probability (SEP) in fading with SEP in the form of \( Q^N(x) \); e.g., differential encoded QPSK where
maximum $N = 4$. The approximation proposed in [10] is integrable in Nakagami-$m$ fading for any $m$. However, it requires summation over all sequences of nonnegative integers $k_1, \ldots, k_m$ such that $k_1 + \cdots + k_m = N$, which is computationally demanding even for small values of $N$. In comparison, (19) is easy to compute for any $N$.

We analyze the average PER of BPSK in Rayleigh fading under the exponential function based approximations of the $Q$-function (Chernoff bound and Wu et al. [8] bound), the upper bound on PER [2] and the proposed approximation in Fig. 3. Figure 3 shows that the proposed approximation and upper bound [2] match the numerical results tightly, and the latter bound is a good candidate for further comparison with (19).

For the BER in (6), the inverse coding gain or threshold, $w_0$, in the upper bound on average PER in Rayleigh fading, i.e., $P_e(\bar{\gamma}) \leq 1 - \exp(-\omega_0/\gamma) [2]$, is approximated in [1] as,

$$\omega_0 \approx k_1 \log N + k_2$$

where $k_1 = 1/k_m$ and $k_2 = (\gamma_c + \log c_m)/k_m$, and $\gamma_c = 0.5772$ is the Euler constant. In [1], for BPSK, the Gaussian $Q$-function is approximated with an exponential function and $k_1$ and $k_2$ are calculated as above. In [3], $\omega_0$ is estimated by fitting the linear model in (24) to the simulations and, $k_1$ and $k_2$ are determined for different modulations. For uncoded 16-QAM, $k_1 = 2.327$ and $k_2 = -3.736$ (see [3], Table I).

In Fig. 4, the approximation error in [1], [3] and (19) for BPSK and 16-QAM in Rayleigh fading is compared. It can be observed that (19) is quite accurate across any value of SNR and packet length. The error for 16-QAM is slightly higher than BPSK, because of the approximations in (9), however this gap reduces quickly with increase in $N$. Figure 4 also shows that (19) is nearly insensitive to SNR and its accuracy mainly depends on the packet length. Whereas, both [1] and [3] yield large errors for small packets even at high SNRs. The errors in these studies decrease with increase in packet length but still remain higher than the proposed method. The approximation (19) performs similarly for FSK, and the approximation error at $N = 256$ is less than 0.04%.

In order to eliminate the effect of inaccuracies introduced by approximations of $w_0$ in [1] and [3], we evaluate $w_0$ numerically and find the average PER from [2]. For 16-QAM in Rayleigh fading, the resulting error is compared with the one in (19) in Fig. 5. It can be seen that the upper bound in [2] is accurate for large values of SNR and packet length while (19) has much better accuracy at low SNRs for any packet lengths. Even at high SNRs, the error in (19) is acceptably small and it reduces sharply with increase in packet length.

VI. CONCLUSIONS

We showed that the PER of uncoded schemes, involving the BER forms of the exponential and the Gaussian $Q$-function, in the AWGN channel can be asymptotically approximated using EVT. Interestingly, the asymptotic distribution of the PER equals the Gumbel distribution function for sample minimum. For uncoded schemes, the EVT approach as compared to the threshold-based bound offers a closed-form and more accurate approximation of the average PER in block-fading channels, while latter approach is still applicable for coded schemes.

REFERENCES


