Delay-Limited Capacity in the Low Power Regime

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Abstract—Outage performance of the \( M \)-block fading with additive white Gaussian noise (BF-AWGN) is investigated in the low-power regime. We consider delay-constrained constant-rate communications with perfect channel state information (CSI) at both the transmitter and the receiver (CSI-TR), under a short-term power constraint (STPC) and a long-term power constraint (LTPC). Subject to STPC, we show that selection diversity that allocates all the power to the strongest block is asymptotically optimal. Then, we provide a simple characterization of the outage probability in the regime of interest. We quantify the reward due to CSI-TR over the constant-rate constant-power scheme and show that this reward increases with the delay constraint. For instance, for Rayleigh fading, we find that a power gain up to 4.3 dB is achievable. Subject to LTPC, we show that the above guidelines still holds and that the outage performance improves due to the flexibility of the LTPC over the STPC. More interestingly, we prove that LTPC allows zero-outage communication even at low SNR and characterize the delay-limited capacity at low SNR in a simple form. More precisely, we establish that the delay-limited capacity scales linearly with the power constraint, for a given \( M < \infty \). Our framework highlights the benefit of fading at low SNR as the delay-limited capacity may outperform the AWGN capacity. For instance, for Rayleigh fading and with \( M = 3 \), the delay-limited capacity is 16\% higher than the capacity of an AWGN channel.

Index Terms—Outage capacity, low-power regime, block-fading channels, channel state information, selection diversity.

I. INTRODUCTION

In many situations, communications should happen in the low-power regime. For instance, wideband communications where the power per sub-band can be very low, sensor network communications, free-space optical communications, etc. The low-power framework is also useful to study communication systems where the power per degrees of freedom (time, space or frequency) is low. A vast body of work exists by now focusing on communications at low-power regime either from performance limits and signaling design perspectives, e.g., [11]–[5], or from more practical views related to deployment, implementation and challenges, e.g., [6], [7], to cite only few. Studying performance limits in the low power regime is also attractive because this regime provides generally the highest energy efficiency defined as the number of bits one can communicate reliably using 1 unit of energy. The latter yardstick is also related to the energy per bit defined as the efficiency defined as the number of bits one is also attractive because this regime provides generally the few. Studying performance limits in the low power regime implementation and challenges, e.g., [6], [7], to cite only

A. Contribution of the paper

In this paper, which is an enhanced version of [23], we consider delay-constrained communications over a block fading (BF) channel with additive white Gaussian noise (AWGN). We assume that communication should happen within a limited number of blocks \( M \) that constitute a frame. Despite the availability of perfect CSI at the transmitter and the receiver (CSI-TR), a constant-rate coding scheme is imposed. With perfect CSI-T, one could adapt the rate to the channel mutual information so that the outage is equal to 0. However, there is an issue with this strategy. In deep fades, the code rate could be very low or even equal to zero, which is not desirable in real-time applications. To circumvent this issue, the notion of capacity versus outage was developed for constant rate real-time applications for a block-fading channel model [19], [24], [25]. In the capacity versus outage problem, a constant target rate aims to be maintained throughout the fading process. An information outage occurs when the target rate is higher than the instantaneous mutual information [25]. However, Power control is exploited to minimize the outage probability. The maximum target rate that can be achieved with an outage probability less than \( \epsilon \) is called the outage capacity, and the zero outage capacity is called the delay-limited capacity in [25]. This problem has been solved in [25]. Nevertheless, as
we argue next (Section II), the solution does not lend itself to nice analytical formulations, thus preventing gaining insight on the impact of system parameters on outage performance. Our focus is on the low-power regime that is formally defined next. We consider first communications under a short-term power constraint (STPC) and then communication under a long-term power constraint (LTPC).

Subject to STPC, we show that the minimum outage takes a simple form. More precisely, we show that for the setting described above, selection diversity is asymptotically optimal. Furthermore, we highlight the value of CSI-TR over CSI-R in terms of outage probability in function of the number of blocks $M$ and for different types of fading. Perhaps the most striking result of this part is the fact that even under a short-term power constraint, the gain in terms of outage probability due to CSI-TR over CSI-R increases unboundedly with $M$, provided that the targeted outage performance is low. Subject to LTPC, we show that the results obtained under STPC still hold and that for an equal power constraint, the LTPC provides a better outage performance than the STPC. Furthermore, we argue that zero-outage communication is possible under LTPC even at low SNR, and we characterize the delay-limited (also called zero-outage) capacity in this regime. Essentially, the capacity is shown to scale linearly with the power constraint. Selection diversity is once again optimal, but our optimal strategy requires channel inversion to achieve zero-outage performance.

We note that our framework, likewise [25], requires CSI-T of the $M$ channel coefficients prior to communication. However, the $M$ blocks may correspond, for instance, to channels in different non-overlapping frequency bands in a multicarrier system with parallel subchannels [25], or may correspond to the number of distinct eigenvalues in a multiple-antenna communication system after a singular value decomposition (SVD) precoding at the transmitter. In that case, having CSI-T of the $M$ fading blocks seems to be feasible. Nevertheless, for large $M$ or/and in case of long term power constraint (LTPC), computing the optimal power profile and the related outage performance becomes cumbersome for an arbitrary (not necessarily small) value of the transmit power. This handicap, however, is avoided in the low power regime since neither the asymptotically optimal power profile under STPC nor the one under LTPC requires all channel gains instantaneously, as we show in the sequel.

B. Outline of the paper

The paper is organized as follows. Section II describes the system model and provide useful related background. In section III, performance limits in the low-power regime under an STPC are derived. The corresponding gain due to CSI-TR over CSI-R only is characterized in section IV. Performance limits under an LTPC are studied in section V. Selected numerical results are presented in section VI. Finally, section VII concludes the paper.

Notations: The logarithms $\log(x)$ is the natural logarithm of $x$. We say that $f(x) \overset{a}{\approx} g(x)$ if and only if $\lim_{x \to a} \frac{f(x)}{g(x)} = 1$. When it is clear from the context, we omit $a$ in $\overset{a}{\approx}$ for convenience. The symbols $\geq$ and $\leq$ are defined analogously. the function $[x]^+$ is equal to $x$ if $x$ is positive and 0 otherwise. $\mathbb{R}_+^M$ designates the $M$-dimensional set of non-negative real numbers. For a non-negative integer $n$, $n!$ represents factorial $n$. $\text{Prob}(\mathcal{A})$ is the probability of the event $\mathcal{A}$. The complement of $\mathcal{A}$ is denoted by $\mathcal{A}^c$. $F^{(k)}(x)$ represent the $n$-derivative of the function $F(\cdot)$ at $x$. Bold face letters represent $M$-dimension vectors, i.e., $\mathbf{x} = (x_1, \ldots, x_M)$.

II. System Model AND Related Background

We consider a block-fading with additive white Gaussian noise (BF-AWGN) channel comprised of $M$ blocks. Each block contains $N$ symbols that undergo the same fading state which is constant, but random. A Codeword of length $M_N = MN$ spans $M$ blocks. The $n$-th, $n = 1, \ldots, N$, channel output at coherence block $m$, $m = 1, \ldots, M$, is expressed by:

$$y(m, n) = h_m x(m, n) + z(m, n),$$

where $x(m, n) \in \mathbb{R}$ is the $n$-th transmitted symbol at time coherence $m$; $h_m \in \mathbb{R}$ is a zero-mean and unit-variance channel gain during block $m$ and $z(m, n) \in \mathbb{R}$ is the AWGN with zero-mean and unity-variance. The fading process $[h_m]$ is assumed to be independent (not necessarily identically distributed) across the $M$ blocks. We assume perfect CSI-TR. For convenience, we let $\gamma_m = |h_m|^2$, $f_m(\cdot)$ its probability density function (PDF), and $F_m(\cdot)$ its cumulative distribution functions (CDF), $m = 1, \ldots, M$. We assume that $F_m(\cdot)$ is defined and infinitely differentiable at 0 to invoke Taylor’s theorem regarding expansion of $F_m(\cdot)$ around 0. We assume that $N$ is sufficiently large ($N \to \infty$) and study performance limits of delay-limited communications at a constant rate $R$, for a given finite number of blocks $M$. Furthermore, the transmitter is constrained by either a short term power constraint (STPC) or a long term power constraint (LTPC) defined, respectively, by [25]:

$$\frac{1}{M} \sum_{m=1}^{M} \mu_m \leq P,$$

$$\frac{1}{K} \sum_{k=1}^{K} \frac{1}{M} \sum_{m=1}^{M} \mu^{(k)}_m \leq P,$$

where $\mu^{(k)}_m = \mathbb{E} \left[ |x^{(k)}(m, n)|^2 \right]$, and the last expectation is with respect to the distribution of the channel input $x^{(k)}(m, n)$, for all $n = 1, \ldots, N$; and $K$ is the length of the $M$-block frames sequence. The maximum mutual information of this channel is given by [25]:

$$I_M(\gamma, \mu) = \frac{1}{M} \sum_{m=1}^{M} \frac{1}{2} \log(1 + \gamma_m \mu_m).$$

The outage probability computed at a rate $R$ is given by:

$$P_{out}^{\text{CSI-TR}}(R, P, M) = \text{Prob}\{I_M(\gamma, \mu) < R\}.$$
only CSI-R is available; we also use the acronym CSI-TR and CSI-R as superscripts on the outage probability to distinguish between the two cases. In case of CSI-TR, \( \mu^{0T} \) and \( \mu^{0R} \) depend on \( \gamma \), i.e., \( \mu^{0T} = \mu^{0T}(\gamma) \) and \( \mu^{0R} = \mu^{0R}(\gamma) \). Then, it is of interest to find the optimal power policy that minimizes the outage probability given by (5).

The minimum outage probability of the BF-AWGN channel has been solved (even in a broader scope) in [25]. We outline briefly the related solution under STPC. Let the region \( Q \) be defined as \( Q = \{ \gamma_h \in \mathbb{R}^M_+ : \gamma_1 \geq \ldots \geq \gamma_M \} \). Assume that \( \gamma \in Q \), then the optimal power over the \( m \)-th block, \( m = 1, \ldots, M \), is given by:

\[
P_m^{0T}(\gamma) = \left[ \lambda^{0T}(b, \gamma) - \frac{1}{\gamma_m} \right]^+, \quad (6)
\]

where \( \lambda^{0T}(b, \gamma) = \frac{1}{b} \sum_{i=1}^{b} \frac{1}{\gamma_i} + \frac{b}{P} \) and \( b \) is the unique integer in \( \{1, \ldots, M\} \) such that \( 1/\gamma_m \leq \lambda^{0T}(b, \gamma) \) for \( m < b + 1 \) and \( 1/\gamma_m > \lambda^{0T}(b, \gamma) \) for \( m \geq b + 1 \). In the previous formulation, \( b \) represents the number of blocks for which a positive instantaneous power is allocated. Indeed, the optimum power profile in (6) has a water-filling interpretation in the sense that a higher channel gain gets more power than a lower one and very poor channels get no power at all. The optimal outage region \( \mathcal{U}(R, P) \) is derived similarly by partitioning \( Q \) into \( M \) sub-regions \( \mathcal{V}_m \subset Q \) such that if \( \gamma \in \mathcal{V}_m \), exactly \( m \) blocks are given positive power. The intersection of \( \mathcal{U}(R, P) \) with \( \mathcal{V}_m, m = 1, \ldots, M \), is given explicitly by:

\[
\mathcal{U}(R, P) \cap \mathcal{V}_m = \left\{ \gamma \in Q : \frac{1}{M} \sum_{i=1}^{m} \frac{1}{\gamma_i} \lambda^{0T}(m, \gamma) < R \right\}. \quad (7)
\]

To extend the optimal power and the outage region given by (6) and (7), respectively, beyond \( Q \) to \( \mathbb{R}^M_+ \), one may sort the component of \( \gamma \) in a non-increasing order, apply the above result to the sorted fading vector and apply the inverse permutation to the obtained power allocation vector and outage probability region.

The minimum outage probability under LTPC can be solved similarly as the one under STPC described above, but using \( s^* \) instead of \( P \), where \( s^* \) is obtained by satisfying the LTPC with equality. Specifically, one can define first the power policy \( \{p^L(\gamma)\} \) by:

\[
p^L_m(\gamma) = \left[ \lambda^L(b, \gamma) - \frac{1}{\gamma_m} \right]^+, \quad (8)
\]

for \( m = 1, \ldots, M \), where \( \lambda^L(b, \gamma) = \left( \frac{\gamma_m}{\sum_{i=1}^{b} \frac{1}{\gamma_i}} \right)^{1/b} \) and where \( b \) is the unique integer in \( \{1, \ldots, M\} \) such that \( 1/\gamma_m \leq \lambda^L(b, \gamma) \) for \( m < b + 1 \) and \( 1/\gamma_m > \lambda^L(b, \gamma) \) for \( m \geq b + 1 \). Then, the optimal power under LTPC is given by:

\[
p^L_m(\gamma) = \begin{cases} p^L(\gamma) & \text{if } \gamma \in \mathcal{R}(s^*) \\ 0 & \text{else,} \end{cases} \quad (9)
\]

where \( s^* \) is defined by:

\[
s^* = \sup \{ s : \mathcal{P}(s) < P \}, \quad (10)
\]

with \( \mathcal{P}(s) \) defined by: \( \mathcal{P}(s) \triangleq \frac{1}{M} \int_{\gamma_{s}}^{\lambda^{L}(1,1)} \left( \frac{1}{M} \sum_{m=1}^{M} p^L_m(\gamma) \right) dF(\gamma) \), and \( \mathcal{R}(s) \) being the region defined by:

\[
\mathcal{R}(s) = \left\{ \gamma \in \mathbb{R}^M_+ : \frac{1}{M} \sum_{m=1}^{M} p^L_m(\gamma) \leq s \right\}. \quad (11)
\]

Clearly, \( \mathcal{R}(s^*) \) in (9) represents the no-outage region. If \( s^* \) in (10) is equal to +\( \infty \), the no-outage region becomes \( \mathcal{R}(s^*) = \mathbb{R}^M_+ \) and the outage probability is equal to 0. This case is of particular interest as it allows characterizing the so-called delay-limited capacity (also called zero-outage capacity) \( C_0(P, M) \) as the solution of:

\[
C_0(P, M) = \arg \min_{R \geq 0} \left\{ \frac{1}{M} \sum_{m=1}^{M} p^L_m(\gamma) = P \right\}. \quad (12)
\]

Note that \( s^* \) may be regarded as an equivalent STPC induced by the LTPC. Intuitively, for a given power \( P \), the minimum outage probability under LTPC offers more flexibility in spending the power than its counterpart under STPC. Hence, it is expected that \( s^* \) be larger than \( P \) which translates into a lower outage probability under LTPC than under STPC.

III. PERFORMANCE LIMITS IN THE LOW-POWER REGIME UNDER A SHORT-TERM POWER CONSTRAINT

From the previous section, it is clear that beyond \( M = 2 \), there is no hope to derive the outage probability in a simple form. The latter difficulty, unfortunately, prevents gaining a useful insight into the behavior of the outage performance with respect to system parameters. We found here another motivation for us to look at the outage performance in the low-power regime.

The low-power regime is captured by letting \( P \to 0 \). In delay unconstrained communication, this is enough since the reliability (arbitrary low decoding probability) is intrinsically contained in the definition of the channel capacity. However, for delay-constrained communications, since the capacity (in Shannon sense) is generally equal to zero, studying the low-power regime can only be meaningful if the targeted outage performance is low in order to guarantee a certain reliability. A simple manipulation of (5) provides:

\[
P^\text{L-CSI-TR}_{\text{out}}(R, P, M) \geq \text{Prob} \left\{ \frac{1}{2} \log \left( 1 + \frac{1}{M} \sum_{m=1}^{M} \gamma_m \mu_m \right) < R \right\} \quad (13)
\]

\[
\geq \text{Prob} \left\{ \frac{1}{2} \log \left( 1 + \frac{1}{M} \sum_{m=1}^{M} \gamma_{\text{max}} \mu_m \right) < R \right\} \quad (14)
\]

\[
\geq \text{Prob} \left\{ \frac{1}{2} \log (1 + \gamma_{\text{max}} P) < R \right\} \quad (15)
\]

\[
= \text{Prob} \left\{ \frac{1}{2} \gamma_{\text{max}} P < R \right\} \quad (16)
\]

\[
= F_{\text{max}} \left( \frac{2R}{P} \right). \quad (17)
\]

\[1\]This might not be the case for a class of fading if CSI-T is available. However, in that case, the outage would be equal to zero regardless.
where (13) follows because \( \frac{1}{M} \sum_{m=1}^{M} \frac{1}{2} \log \left( 1 + \gamma_m \mu_m \right) \leq \frac{1}{2} \log \left( 1 + \frac{1}{M} \sum_{m=1}^{M} \gamma_m \mu_m \right) \) due to the Jensen’s inequality; (14) holds since \( \gamma_{\text{max}} = \max \gamma_m \); (15) is a consequence of the STPC; to obtain (16), we used the inequality \( \log (1 + x) \leq x \) and finally \( F_{\text{max}}(\cdot) \) in (17) is the CDF of \( \gamma_{\text{max}} \). From (17), it is clear that if \( P \to 0 \) and \( R \) is arbitrary (not necessarily small), then the outage probability is arbitrary close to 1. Therefore, in delay-constrained communications, and under STPC, reliability at low-power regime can only be obtained if \( R \to 0 \). Hence, with STPC, the low-power regime we are interested in corresponds to the situation where \( P, R \) and the outage performance tend all toward 0. We are now ready to state our main result.

**Theorem 1:** For the block-fading channel described by (1), with perfect CSI-TR, and subject to STPC (2), the outage performance is characterized as follows:

\[
P_{\text{out}}(R, P, M) \approx \left\{ \begin{array}{ll} 
M P & \text{if } \gamma_m = \max_{1 \leq k \leq M} \gamma_k \\
0 & \text{otherwise} 
\end{array} \right.
\]

The corresponding outage probability is equal to

\[
P_{\text{out}}(R, P, M) = \text{Prob} \left( \frac{1}{2M} \log (1 + M P \gamma_{\text{max}}) < R \right) = F_{\text{max}}(x_1)
\]

\[
x_1 \approx F_{\text{max}} \left( 2R \over P \right)
\]

where \( x_1 \) in (21) is equal to \( x_1 = \frac{2R P}{PM} \) and where the asymptotic equality (22) holds true since \( R \to 0 \) (please see Appendix A). Hence, the lower bound is asymptotically achievable, and Theorem 1 is thus proved.

Few remarks are worthwhile.

**Remark 1:** For the class of fading such that \( F_m(x) > 0 \) for all \( x > 0, m = 1, \ldots, M \) (this is the case of most popular fading in wireless communications), since \( F_{\text{out}}(R, P, M) \to 0 \), then Theorem 1 implies that we necessarily have \( \frac{R}{P} \to 0 \). That is, for this class of fading, reliable communication (communication with arbitrary low outage probability) at low-power regime requires not only that the rate \( R \) be small (\( R \to 0 \)), but more stringently that the ratio \( \frac{R}{P} \) be small too (\( \frac{R}{P} \to 0 \)).

**Remark 2:** While it is well-known that increasing \( M \) can only improve outage performance (assuming optimal coding), Theorem 1 describes how the outage probability decreases with \( M \). In fact, if the number of blocks over which communication is happening increases from \( M \) to \( M + 1 \), the outage probability decreases by a factor of \( F_{M+1}(\frac{2R}{P}) \). More specifically, one can express the outage probability in a more explicit way as follows:

\[
\log \left( P_{\text{out}}(R, P, M) \right) \approx \log \left( \prod_{m=1}^{M} F_{m}(2R \over P) \right) = \log \left( \prod_{m=1}^{M} \frac{F_{m}(0)}{n_m!} \left( \frac{2R}{P} \right)^{n_m} \right) = -\log \left( \frac{P}{2R} \right) \sum_{m=1}^{M} n_m
\]

where \( n_m \) is the smallest integer such that \( F_{m}(0) \neq 0, m = 1, \ldots, M \); (24) is obtained by series expansion of \( F_m(x) \) around zero and because \( \frac{R}{P} \to 0 \) (see Remark 1) and (25) follows because \( \lim_{\frac{R}{P} \to 0} \log \left( \frac{F_{m}(0)}{n_m!} \right) = 0 \). Since in (24) Taylor expansion is performed to the \( n_m \)th order, \( m = 1, \ldots, M \), then the error incurred by such an approximation goes faster to 0 than \( \left( \frac{2R}{P} \right)^{\sum_{m=1}^{M} n_m} \) as \( \frac{R}{P} \to 0 \). For fixed \( P \) and \( R \), (26) describes how the outage probability depends on \( M \). Note that since \( n_m \geq 1 \), for all \( m = 1, \ldots, M \), then \( \sum_{m=1}^{M} n_m \geq M \). Also, because \( \frac{P}{2R} \to \infty \), then \( \sum_{m=1}^{M} n_m \) may be regarded as the diversity order provided by coding over \( M \) blocks at low-power regime.

**Remark 3:** The result in Theorem 1 translates directly to a characterization of the so-called outage capacity. Indeed, along similar steps as Remark 2, it is easy to verify that the outage capacity \( C_\epsilon(\cdot) \) is given by:

\[
C_\epsilon(P, M) \approx \frac{1}{2} \left( \prod_{m=1}^{M} \frac{n_m!}{F_m(0)} \right)^{\frac{1}{\epsilon}} \epsilon^{\frac{1}{\epsilon n_m} P}
\]

The ratio \( C_{\epsilon}(P, M) \), defined as the capacity per unit power for a given outage probability \( \epsilon \), is a key performance measure in the regime of interest [21]. Therefore, our result also characterizes the capacity per unit power through (27).

We conclude this section by mentioning that although, in the low-power regime, channel estimation might be challenging, the optimal scheme given by (19) does not perform power adaptation. Indeed, the optimal scheme only requires few feedback bits in each frame of \( M \) blocks to designate the block with the highest channel gain. More precisely, it needs exactly \( \lceil \log_2(M) \rceil \) feedback bits, where \( \lceil \cdot \rceil \) is the ceiling function. A question of interest, in this case, would be to assess whether the additional gain provided by CSI-TR over CSI-R is worth the above low rate feedback mechanism. This is answered in Section IV.

Note that we are computing the outage probability in the log scale in order to show the diversity order. Also note that although our focus is on low-power regime, the diversity order is still meaningful since the outage probability is arbitrary driven to zero as a consequence of the fact that \( \frac{R}{P} \to 0 \).
IV. HOW REWARDING THE OUTAGE PROBABILITY AT THE TRANSMITTER IS?

We first evaluate the perfect outage probability assuming perfect CSI-R and no CSI-T. Recall that with CSI-R and under an STPC, the outage probability (using constant power over the blocks) is given by [24]:

$$P_{out}^{CSI-R}(R,P,M) = \text{Prob} \left\{ \frac{1}{M} \sum_{m=1}^{M} \log \left( 1 + P \gamma_m \right) < R \right\}$$

(28)

$$\approx \text{Prob} \left\{ \sum_{m=1}^{M} \gamma_m < \frac{2M R}{P} \right\}$$

(29)

$$= F_{\gamma} \left( \frac{2M R}{P} \right),$$

(30)

where (29) follows because $\log(1+x) \approx x$ around zero, and where $F_{\gamma}(\cdot)$ is the CDF of $\sum_{m=1}^{M} \gamma_m$. Note that the right hand side (RHS) of (29) is asymptotically equal to the outage probability of a repetition diversity with constant-power transmission. Hence, with CSI-R, repetition diversity with constant-power transmission is an optimal strategy at asymptotically low-power regime.

There are no explicit expression of $F_{\gamma}(\cdot)$ for general fading distributions of $\gamma_m$, $m = 1, \ldots, M$. Nevertheless, we have the following result that characterizes the gain in terms of outage performance due to CSI-TR.

**Theorem 2:** Let $n_m$ be the smallest integer such that $F_{\gamma_m}(0) \neq 0$, $m = 1, \ldots, M$. Then, we have:

$$\lim_{P \to 0} \frac{P_{out}^{STPC}(R,P,M)}{P_{out}^{CSI-R}(R,P,M)} = \frac{\left( \sum_{m=1}^{M} n_m \right)!}{M^{\sum_{m=1}^{M} n_m} \prod_{m=1}^{M} M^{n_m}!}.$$

(31)

**Proof:** The proof is presented in Appendix B.

Since the availability of CSI-T can only improve outage performance $P_{out}^{STPC}(R,P,M)$, one needs to verify that the right-hand side (RHS) of (31) is at most equal to 1. This is not very clear at first glance. However, we show below that this is, in fact, a direct consequence of the following lemma that provides a more general result.

**Lemma 1:** Let $n$ be a positive integer. Let $V_n$ be the simplex defined by:

$$V_n = \left\{ (t_1, \ldots, t_n) \in [0, \infty)^n : \sum_{k=1}^{n} t_k \leq 1 \right\}.$$

Then, for any $(a_1, \ldots, a_n) \in (0, \infty)^n$, we have:

$$\frac{\Gamma(1 + a_1 + \cdots + a_n)}{\Gamma(1 + a_1) \cdots \Gamma(1 + a_n)} \leq \frac{\sum_{k=1}^{n} a_k}{n},$$

(32)

where $\Gamma(\cdot)$ is the Gamma function defined by: $\Gamma(x) = \int_{0}^{\infty} e^{-t} t^{x-1} dt$, for $x > 0$.

**Proof:** We first recall a multidimensional integral property of the Gamma function [26]:

$$\int_{V_n} t_1^{a_1-1} \cdots t_n^{a_n-1} dt_1 \cdots dt_n = \frac{\Gamma(a_1) \cdots \Gamma(a_n)}{\Gamma(1 + a_1 + \cdots + a_n)}$$

(33)

Then, we need to lower bound the left hand side (LHS) of (33). For this purpose, we let $X_k$ be a random variable with PDF $f_X(t) = a_k t^{a_k-1}$, for $t \in [0,1]$, $k = 1, \ldots, n$.\(^3\) We also assume that $X_k$’s are jointly independent such that their joint PDF $f_{X_1,\ldots,X_n}(t_1,\ldots,t_n) = \prod_{k=1}^{n} f_X(t_k)$.

The LHS of (33) can be lower bounded as follows:

$$\int_{V_n} t_1^{a_1-1} \cdots t_n^{a_n-1} dt_1 \cdots dt_n$$

$$= \frac{1}{\prod_{k=1}^{n} a_k} \int_{V_n} \prod_{k=1}^{n} f_X(t_k) dt_1 \cdots dt_n$$

$$= \frac{1}{\prod_{k=1}^{n} a_k} \int_{V_n} \prod_{k=1}^{n} f_{X_1,\ldots,X_n}(t_1,\ldots,t_n) dt_1 \cdots dt_n$$

$$= \frac{1}{\prod_{k=1}^{n} a_k} \text{Prob} \left\{ \sum_{k=1}^{n} X_k \leq 1 \right\}$$

$$\geq \frac{1}{\prod_{k=1}^{n} a_k} \text{Prob} \left\{ X_k \leq \frac{1}{n}, k = 1, \ldots, n \right\}$$

(37)

$$= \frac{1}{\prod_{k=1}^{n} a_k n^a}.$$  

(38)

Since for any $x > 0$, $\Gamma(1 + x) = x \Gamma(x)$, (32) follows from (33) and (38) immediately.

Finally, we note that the RHS of (31) is indeed smaller than one because:

$$\frac{1}{M^{\sum_{m=1}^{M} n_m}} \prod_{m=1}^{M} \frac{\Gamma(1 + n_1 + \cdots + n_M)}{M^{n_1} \cdots M^{n_M}} \leq 1,$$

(39)

where the last inequality follows from Lemma 1.

There is no particular difficulty in evaluating the RHS of (31) for given fading channels. For example, for Rayleigh fading, $n_m = 1$, $m = 1, \ldots, M$, and the RHS of (31) is equal to $M^M$. The later result holds for any fading channel that satisfies $n_m = 1$, $m = 0, \ldots, M - 1$. Since $\frac{M^M}{n^M}$ is strictly decreasing in $M$, then it follows from Theorem 2 that for this class of fading, increasing $M$ strictly increases the gap between $P_{out}^{STPC}(R,P,M)$ and $P_{out}^{CSI-R}(R,P,M)$. More generally, for all fading channels such that $n_m$ is constant, $m = 1, \ldots, M$, we have the following.

**Corollary 1:** Let $n_m$ be the smallest integer such that $F_{\gamma_m}(0) \neq 0$. Then, if $\forall m = 1, \ldots, M$, $n_m = n$, then we have:

i. $$\lim_{P \to 0} \frac{P_{out}^{STPC}(R,P,M)}{P_{out}^{CSI-R}(R,P,M)} = \frac{(Mn)!}{M^M (n!)^M}.$$  

(40)

ii. the ratio $\frac{(Mn)!}{M^M (n!)^M}$ is strictly decreasing in $M$, for any $n \geq 1$ and thus increasing $M$ strictly enhances the advantage of CSI-TR over CSI-R, in term of outage probability.

iii. the ratio $\frac{(Mn)!}{M^M (n!)^M}$ converges to 0 as $M \to \infty$ and hence the advantage of CSI-TR over CSI-R is unbounded.

**Proof:** Property i in Corollary 1 is an immediate consequence of Theorem 2 and the fact that $n_m = n$, for all $m = 0, \ldots, M - 1$. To prove ii, we recall first the Stirling’s formula [27]:

$$1 \leq \frac{k!}{\sqrt{2\pi k}} \left( \frac{k}{e} \right)^k \leq \frac{e}{\sqrt{2\pi}}.$$  

(41)

\(^3\)This is because $\frac{(Mn)!}{M^M (n!)^M}$ is strictly decreasing in $M$, for any $n \geq 1$ and thus increasing $M$ strictly enhances the advantage of CSI-TR over CSI-R, in term of outage probability.
for any integer \( k \geq 1 \). Then, we let \( U(M) = \frac{(Mn)!}{M^M (n!)^M} \) and use (41) to upper bound \( U(M) \) as follows:

\[
1 \leq \frac{U(M)}{\left(\frac{e}{\sqrt{2\pi}}\right)^{Mn} \sqrt{2\pi Mn}} \leq e^{-\frac{1}{2\pi}}.
\]  

(42)

Now, with the help of (42), one can upper bound \( \frac{U(M+1)}{U(M)} \) as follows:

\[
\frac{U(M+1)}{U(M)} \leq e^{\frac{(M+1)n}{\sqrt{2\pi}}} \sqrt{\frac{2\pi(M+1)n}{(n!)^{M+1}}} \left(\frac{1}{\sqrt{2\pin}}\right)^{Mn}.
\]  

(43)

\[
= e^{-\frac{1}{\sqrt{2\pi}}} \frac{(M+1)n}{\sqrt{2\pin}} \leq e^{-\frac{1}{\sqrt{2\pi}}} \frac{M+1}{M} \sqrt{\frac{M+1}{n!}}.
\]  

(44)

\[
\leq e^{-\frac{1}{\sqrt{2\pi}}} \frac{M+1}{M} \sqrt{\frac{M}{n!}}.
\]  

(45)

\[
\leq e^{-\frac{1}{\sqrt{2\pi}}} \frac{M+1}{M} \sqrt{\frac{M}{n!}}.
\]  

(46)

\[
< 1.
\]  

(47)

\[
\leq e^{-\frac{1}{\sqrt{2\pi}}} \frac{M+1}{M} \sqrt{\frac{M}{n!}}.
\]  

(48)

where (45) is obtained using the LHS of (41); (46) holds because \( n \geq 1 \) and (47) follows from the fact that the function \( x \mapsto \sqrt{\frac{x+1}{x}} \) is monotonically decreasing and thus for all \( M \geq 1 \), \( \sqrt{\frac{M+1}{M}} \leq \sqrt{\frac{M+1}{M}} = \sqrt{M} \). Therefore, \( U(M) \) is strictly increasing in \( M \) and property iii is thus proved. Finally, in order to verify iii, we again use (42) and observe that:

\[
U(M) \leq e^{\frac{(Mn)!}{\sqrt{2\pin}}} \sqrt{\frac{2\pi Mn}{(n!)^{M}}}.
\]  

(49)

\[
= e^{\left(\frac{n}{\sqrt{2\pin}}\right)^M} \sqrt{n!} \sqrt{M}.
\]  

(50)

\[
\leq e^{\frac{1}{\sqrt{2\pin}}} \sqrt{n!} \sqrt{M}.
\]  

(51)

where (51) follows from the LHS of (41). Clearly, the RHS of (51) converges to zero as \( M \to \infty \) and since \( U(M) \) is non-negative, then \( \lim_{M \to \infty} U(M) = 0 \). Hence, property iii holds true and Corollary 1 is proved.

Note that Corollary 1 encompasses the important particular case where the fading over the \( M \) blocks are identically distributed since, in that case, \( n_m \) would be constant for all \( m = 1, \ldots, M \). Furthermore, in light of Corollary 1, it can be inferred that the advantage of CSI-TR over CSI-R can be tremendous for \( M \) sufficiently large, at low-power regime, provided that the targeted outage performance (or equivalently the rate \( R \) at which we communicate) is low. For instance, using (18), the power required to achieve an outage performance \( \epsilon \) while communicating at a rate \( R \) assuming CSI-TR is asymptotically equal to:

\[
\text{Pout,CSI-TR (\( \epsilon, R \))} \approx \frac{2R}{\left(\prod_{m=1}^{M} \epsilon_{\frac{n_m}{F_{m}^{(0)}}}^{n_m} \right)^{\frac{1}{\epsilon_{\frac{n_m}{F_{m}^{(0)}}}^{n_m}}}}.
\]  

(52)

Whereas with CSI-R only, the required power is equal to:

\[
\text{Pout,CSI-R (\( \epsilon, R \))} \approx \frac{2MR}{\left(\sum_{m=1}^{M} \frac{n_m}{F_{m}^{(0)}}\right)^{\frac{1}{\epsilon_{\frac{n_m}{F_{m}^{(0)}}}^{n_m}}}}.
\]  

(53)

Hence, the power gain due to CSI-TR is equal to:

\[
\frac{\text{Pout,CSI-TR (\( \epsilon, R \))}}{\text{Pout,CSI-R (\( \epsilon, R \))}} \approx \left(\frac{\sum_{m=1}^{M} \frac{n_m}{F_{m}^{(0)}}}{\prod_{m=1}^{M} \frac{n_m}{F_{m}^{(0)}}}\right)^{\frac{1}{\epsilon_{\frac{n_m}{F_{m}^{(0)}}}^{n_m}}}.
\]  

(54)

Now, if \( M = 2 \) and \( n_m = 1 \), for all \( m = 1, \ldots, M \), (54) yields 1.5 dB gain which is considerable in the low-power regime. Increasing \( M \) to 3 enhances further the gain to about 2.18 dB. Actually, it can be easily shown that as \( M \to \infty \), the ratio \( \frac{\text{Pout,CSI-TR (\( \epsilon, R \))}}{\text{Pout,CSI-R (\( \epsilon, R \))}} \) converges to \( e^{-1} \) and hence for \( M \) sufficiently large and assuming all \( n_m = 1 \), a remarkable 4.3 dB gain is achievable. Note that there is no contradiction between the fact that the advantage of CSI-TR over CSI-R in terms of outage probability is unbounded as stated by iii in Corollary 1 and the fact that the power gain resulting from CSI-T is bounded. This is because the exponent in (54) also depends on \( M \). Therefore, the common belief by which “optimal power allocation under constant power” is somehow misleading. In [25], the authors pointed out that for Rayleigh fading channels, “As \( M \) increases, the outage probability with the short-term constraint is always very close to that for constant power”. This guideline is unfortunately not very precise and hence should be looked at more carefully, at least in the low-power regime. Indeed, as stated by Corollary 1, the outage probability with the short-term constraint can be arbitrary smaller than that of constant power, for large \( M \).

So far, we have focused on the case where all \( n_m \) are equal, i.e., \( n_m = n \) is constant and fixed, and we have analyzed the value of CSI-TR over CSI-R in function of \( M \). Another aspect of interest would be to fix \( M \), and study the reward provided by CSI-TR versus \( n \). This would help understanding the value of CSI-TR for different fading channels, for a given delay constraint. For convenience, our result is summarized in the following corollary.

**Corollary 2:** Let \( n_m \) be the smallest integer such that \( F_{m}^{(0)}(0) \neq 0 \), \( m = 1, \ldots, M \). Then, if \( \forall m = 1, \ldots, M, n_m = n \), then we have:

i. the ratio \( \frac{(Mn)!}{M^M (n!)^M} \) is strictly decreasing in \( n \), for any \( M > 1 \), and thus the higher \( n \), is the more important is the advantage of CSI-TR over CSI-R, in term of outage probability.

\[
\lim_{n \to \infty} \frac{(Mn)!}{M^M (n!)^M} = \begin{cases} 
0, & \text{if } M > 1 \\
1, & \text{if } M = 1.
\end{cases}
\]  

(55)

**Proof:** First, we note that if \( M = 1 \), the ratio in Corollary 2 is equal to 1, irrespective of \( n \). Now, assume that \( M > 1 \) and...
let \( V(n) = \frac{(Mn)!}{n^M (e^n)^n} \). Since the bounds in (42) hold for \( V(n) \), then we can upper bound \( \frac{V(n+1)}{V(n)} \) as follows:

\[
\frac{V(n+1)}{V(n)} \leq \frac{e^{(n+1)} \left( \frac{n+1}{e} \right)^{M(n+1)} \sqrt{2\pi nM(n+1)} \left( \frac{n}{e} \right)^M}{((n+1))!} \frac{1}{e^{2\pi M n(n+1)} \left( \frac{n}{e} \right)^M} = \frac{1}{\sqrt{2\pi} e^M} \leq \frac{e}{\sqrt{2\pi} e^M} \sqrt{\frac{1+n}{n}} \leq \frac{e}{\sqrt{2\pi} e^M} \sqrt{\frac{2M^2}{(2n)^M}} \leq \frac{e}{\sqrt{2\pi} e^M} \left( \frac{2+\frac{1}{2}}{2} \right) \log(2) - x \leq \frac{1}{\sqrt{2\pi} e^M} < 1,
\]

(58)

where (58) follows because the function \( x \mapsto \sqrt{\frac{1+n}{x}} \left( \frac{1+n}{x} \right)^M \) is the product of two positive-definite monotonically decreasing functions and thus is also monotonically decreasing; (60) is true because the function \( x \mapsto (x + \frac{1}{2}) \log(2) - x \) is monotonically decreasing and also because \( M \geq 2 \). Hence \( V(n) \) is strictly decreasing in \( n \). To prove ii, we upper bound \( V(n) \) departing from (49) as follows:

\[
\frac{V(n)}{V(n)} \leq \frac{e}{\sqrt{2\pi} e^M} \left( \frac{n}{e} \right)^M \sqrt{\frac{2\pi Mn}{(n!)^M}} \leq \frac{e}{\sqrt{2\pi} e^M} \frac{(n/e)^M}{\sqrt{2\pi n} \left( \frac{e}{2} \right)^n} = \frac{e}{\sqrt{2\pi} e^M} \leq \frac{1}{\sqrt{2\pi} e^M} \leq \frac{1}{\sqrt{2\pi} e^M} \leq \frac{1}{\sqrt{2\pi} e^M} \left( \frac{2+\frac{1}{2}}{2} \right) \log(2) - x \leq \frac{1}{\sqrt{2\pi} e^M} < 1,
\]

(64)

where (64) is due to the LHS of (41). Since the RHS of (65) converges to \( 0 \) as \( n \to \infty \) for any \( M > 1 \), then so does \( V(n) \). This completes the proof of Corollary 2.

\[ \]

V. PERFORMANCE LIMITS IN THE LOW-POWER REGIME UNDER A LONG-TERM POWER CONSTRAINT

As we highlighted at the end of Section II, the minimum outage probability under LTPC may be derived in a similar way as its counterpart under STPC, but using \( s^* \) instead of \( P \). However, in order for our results in the low power regime to apply under LTPC, one need to verify that when it exists and is finite, \( s^* \) → 0 as \( P \to 0 \), i.e., \( \lim_{P \to 0} s^*(P) = 0 \). This is true as claimed by Lemma 2.

**Lemma 2:** Define the function \( \Psi(P, R) = \frac{1}{M} \sum_{m=1}^{M} \gamma_m \). For the minimum outage probability of block fading channels described by (1) under LTPC (3), the equivalent STPC \( \Psi(P, R, \gamma_m)(P) \) defined by (10) is either infinite or converges to 0 as \( P \to 0 \). More precisely:

- If \( \exists a > 0 \) such that \( \Psi(P, R, \gamma_m)(P) \leq 0, \forall P \in [0, a], \) then \( \lim_{P \to 0} s^*(P, R, \gamma_m)(P) = +\infty \).
- else, \( \lim_{P \to 0} s^*(P, R, \gamma_m)(P) = 0 \).

\[ \]

**Proof:** For convenience, the proof is presented in Appendix C.

We note that \( \frac{1}{M} \sum_{m=1}^{M} \rho_m(\gamma) \) in Lemma 2 depends on \( R \), which itself may scale with the power budget \( P \). The case where \( R \) does not scale with \( P \) may be regarded as a particular case where \( R(P) = R \), and thus the first case in Lemma 2 cannot happen. In that case, it holds that \( \lim_{P \to 0} s^*(P, R) = 0 \) by Lemma 2. On the other hand, if \( R \) scales with \( P \) such that \( \Theta(P, R, \gamma_m) \leq 0, \forall P \in [0, a], \) then the equivalent STPC \( \Psi(P, R, \gamma_m)(P) \) induced by the LTPC \( P \) tends toward infinity. As we pointed out in Section II, the latter situation depicts a zero-outage performance itself characterized by delay-limited capacity. We are now ready to state the following result.

**Theorem 3:** For the block-fading channel described by (1), with perfect CSI at both the transmitter and the receiver, and subject to LTPC (3), the outage performance is characterized as follows:

- If \( \exists a > 0 \) such that \( \Theta(P, R, \gamma_m) \leq 0, \forall P \in [0, a], \) then \( \lim_{P \to 0} P_{out}^{\text{CSI-TR}}(R(P), P, M) = 0 \),
- else, \( \lim_{P \to 0} P_{out}^{\text{CSI-TR}}(R(P), P, M) \approx \prod_{m=1}^{M} F_m(y_1) \),

where \( y_1 = \frac{\gamma_m}{M s^*} \), where \( s^* \) is the solution of (10).

**Proof:** First, assume there exists no \( a > 0 \) such that \( \Theta(P, R, \gamma_m) \leq 0, \forall P \in [0, a] \). Then by Lemma 2, we have that \( s^* \) converges to 0 as \( P \to 0 \). In this case, along similar lines as in the case of STPC, the lower bound (17) holds true, the only difference is that \( P \) is now replaced by \( s^* \). Thus, we have:

\[ P_{out}^{\text{PT}}(R, P, M) \geq \prod_{m=1}^{M} F_m \left( \frac{2R}{s^*} \right) \]

(66)

Next, we provide a power policy that satisfies the LTPC and achieves asymptotically the lower bound in (66). For this purpose, let us consider the following power profile:

\[ P_{m,l}^{\text{PT}}(\gamma) = \begin{cases} \frac{e^{\gamma M - 1}}{\gamma M s^*} & \text{if } \gamma_m = \frac{\gamma_k}{M s^*} \text{ and } \gamma_m > \frac{e^{\gamma M - 1}}{\gamma M s^*} \\ 0 & \text{otherwise} \end{cases} \]

(67)

Clearly the power policy given by (67) fulfills the LTPC since:

\[ E \left[ \frac{1}{M} \sum_{m=1}^{M} P_{m,l}^{\text{PT}} \right] = E \left[ \frac{1}{M} \sum_{m=1}^{M} \left( \frac{\gamma M - 1}{\gamma M s^*} - 1 \right) \right] = \frac{1}{M} \sum_{m=1}^{M} \left( \frac{\gamma M - 1}{\gamma M s^*} - 1 \right) = P, \]

(69)

where (69) holds true since (68) only accounts for the power spent over 1 (the strongest) block. Furthermore, this power policy achieves a probability of outage equal to \( \text{Prob} \left[ \gamma_{\max} \leq \frac{e^{\gamma M - 1}}{\gamma M s^*} \right] \), which is asymptotically equal to the lower bound in (66).

Second, assume that \( \Theta(P, R, \gamma_m) \leq 0 \) holds on an interval of the form \([0, a] \), with \( a > 0 \), then, again, by Lemma 2, we

\[ \]
have that \( s^*(P, R(P)) \) tends toward \(+\infty\) as \( P \to 0 \). It follows that the no-outage region \( \mathcal{R}(s^*) \) defined by (11) is equal to \( \mathbb{R}_+^M \) which yields an outage probability asymptotically equal to 0. This completes the proof of Theorem 2.

We note that in the case where \( \lim_{P \to 0} s^* = 0 \), we still have that \( s^* \geq P \), i.e., \( s^* \) converges to 0 slower than \( P \). Intuitively, for equal power constraint, an LTPC provides more flexibility than an STPC and hence better performances. Since for bounded \( s^* \), outage performance under LTPC is similar to that under STPC, but using \( s^* \) instead of \( P \), as asserted by Theorem 3, then one expects \( s^* \) to be larger than \( P \). This can be confirmed as follows. Recalling (9), we have:

\[
P = \int_{\mathcal{R}(s^*)} \left( \frac{1}{M} \sum_{m=1}^{M} \mu_m^s(\gamma) \right) dF(\gamma) \tag{71}
\]

\[
\leq \int_{\mathcal{R}(s^*)} s^* dF(\gamma) \tag{72}
\]

\[
=s^* \int_{\mathcal{R}(s^*)} dF(\gamma) \tag{73}
\]

\[
\leq s^*, \tag{74}
\]

where (72) follows from the definition of \( \mathcal{R}(s) \) in (11), and (74) holds true since the integral in (73) is a probability.

Remark 4: In the case where \( s^* \to 0 \) as \( P \to 0 \), all the results established in Sections III and IV under STPC apply to the LTPC setting, but substituting \( P \) by \( s^* \).

In regard of Lemma 2 and (74), not only \( s^* \) is larger than \( P \), but it may converge asymptotically to infinity as \( P \to 0 \). In this case, one can achieve zero-outage performance over a finite number of blocks \( M \) while communication at a rate arbitrary close to the delay-limited capacity. The next theorem is an asymptotic characterization of the delay-limited capacity.

Theorem 4: For the block-fading channel described by (1), with perfect CSI at both the transmitter and the receiver, and subject to LTPC (3), the delay-limited capacity \( C_0(P, M) \) is given by:

\[
C_0(P, M) \approx \frac{P}{2E[\gamma_{\max}^{-1}]} \tag{75}
\]

provided that \( E[\gamma_{\max}^{-1}] \) is finite. Otherwise, the delay limited capacity is equal to zero.

Proof: Assume that \( E[\gamma_{\max}^{-1}] \) is finite. Otherwise, we split the proof into two parts. A converse part where we show that any zero-outage achievable rate is asymptotically smaller than the RHS of (75). Then, we prove the achievability part.

a) Converse: For a given (asymptotically small) \( P \) and a given outage performance \( \epsilon \), let \( R \) be the rate, such that the minimum outage probability under LTPC \( P \), is equal to \( \epsilon \). We know from Lemma 2 that \( s^* \), the correspondent equivalent STPC as defined by (10), either converges to infinity or to 0, as \( P \to 0 \). Clearly, the latter case induces a no outage region asymptotically equal to the origin which translates into \( R = 0 \) and thus only the former case is of interest. It is easy to verify that the former case happens if and only if \( \Theta(P, R) \leq 0 \). Hence, \( \exists a > 0 : \forall P \in [0, a) \), we have:

\[
P \geq \frac{P}{2E[\gamma_{\max}^{-1}]} \sum_{m=1}^{M} p_m^s(\gamma) \tag{76}
\]

Figure 1. Outage probability versus SNR for Rayleigh fading channels, for different \( M \) values, and \( R = 10^{-4} \) ncpu, in the low-power regime.

where \( \exists a > 0 : \forall P \in [0, a) \), we have:

\[
P \geq \frac{P}{2E[\gamma_{\max}^{-1}]} \sum_{m=1}^{M} p_m^s(\gamma) \tag{76}
\]
the curves are very low as they correspond to normalized performance measures per degrees of freedom. For instance, in ultra-wideband applications, the bandwidth available is of order of five hundreds mega Hertz (500MHZ). This translates a normalized rate per degrees of freedom $R = 10^{-4}$ bits per second per Hertz (b/s/Hz) into an aggregate rate of 50 Kb/s which is decently high for low-rate applications such as sensor data collection, precision locating and tracking applications. Figure 1 shows the accuracy of our asymptotic analysis since the exact and the asymptotic representations match perfectly. As $M$ increases, the ratio between the outage probabilities under CSI-TR and CSI-R decreases as predicted by Corollary 1. Also, for a given outage performance, the power gain due to CSI-T increases from 0 dB for $M = 1$ to about 1.5 dB for $M = 2$ and reaches 2.18 dB for $M = 3$, exactly as predicted by (54). Finally, we note that the slopes of the curves in Fig. 1 are equal to 1, 2 and 3 for $M = 1$, $M = 2$ and $M = 3$, respectively, since for Rayleigh fading, $n_m = 1$, as discussed in Remark 2. For Nakagami-$m$ fading, in addition to the above observations, we note first that the slopes of the curves in Fig. 2 are equal to 2, 4 and 6 for $M = 1$, $M = 2$ and $M = 3$, respectively, since for Nakagami-$m$ ($m = 2$) fading, $n_m = 2$. Comparing the impact of the two fading on the outage, we observe that for $M = 3$ and $P = -4$ dB, for instance, the ratio between the outage probabilities with CSI-TR and with CSI-R is equal to about 0.22 in Fig. 1 whereas it is equal to 0.12 in Fig. 2, implying that the advantage of CSI-TR increases with $n_m$, again in agreement with Corollary 2. We note that the asymptotic characterization in Theorem 1 holds true as long as $R \to 0$, irrespective of the SNR values. The fact that the rate $R$ is set very small in both figures 1 and 2 justifies why the asymptotic curves and the ones obtained by simulations match perfectly well. Should $R$ be chosen not very small, for example $R = 0.1$ nats per channel use (npcu), we expect to see a discrepancy between the asymptotic curves and the ones obtained by simulations. This fact is confirmed by Fig. 3. Second, come our results for the LTTPC. Figure 4 depicts the outage performance for Rayleigh fading under STPC and LTTPC, with $M = 2$. In Fig. 4, the rate is set to $R = 0.01$ nats per channel use. Beyond the accuracy of our asymptotic analysis (in Theorems 1 and 3) which curves match perfectly with the exact ones, Fig. 4 highlights the tremendous gain provided by LTTPC over STPC even at low-SNR. For instance, at an outage performance equal to $10^{-3}$, the power gain is about 14 dB. Of particular interest, is the power corresponding to the delay-limited capacity plotted with a vertical line at $SNR = -15.48$ dB. Indeed, this is the power needed to communicate $R = 0.01$ nats per channel use (npcu) with zero-outage, again in full agreement with Theorem 4. In order to verify the increase of the delay-limited capacity with $M$, we depict in Fig. 5 the delay limited capacity versus $SNR$ curves for $M = 2$ and $M = 3$. Recall that for $M = 1$, the delay limited capacity is equal to 0 for Rayleigh fading channels. Likewise, with CSI-R only, the delay limited capacity is also equal to zero. In Fig. 5, the asymptotic curves have been obtained using Theorem 4 whereas the exact curve for $M = 2$ has been obtained by solving numerically the following equation [25]:

$$P = \left(e^{4C_0}-1\right)\log\left(1+e^{-4C_0}\right)+e^{2C_0}(\pi-4\arctan(e^{-2C_0}))-4C_0.$$  

The curve corresponding to exact delay-limited capacity for $M = 3$ has been obtained by solving numerically $\Theta(P,C_0) = 0$, where $\Theta(\cdot,\cdot)$ is defined in Lemma 2. As can be seen in Fig. 5, the exact curves and the asymptotic ones match perfectly at low SNR. As SNR increases (above -10 dB for $M = 2$ and above -15 dB for $M = 3$), the exact and the asymptotic curves are more distinguishable which suggests that the asymptotic analysis does not predict perfectly well the delay-limited capacity behavior in these SNR ranges. However, both the exact curves and the asymptotic ones confirm the capacity gain due to an increase of the frame length $M$. Indeed, increasing $M$ from 2 to 3 provides 60 % capacity increase in the low-power regime. More interestingly, the delay-limited capacity for $M = 3$ is higher than the AWGN capacity at low SNR. This highlights, once again, the benefit of fading in the low-power regime, even under stringent delay constraints. Finally, we note that as $M \to \infty$, we expect the delay-limited capacity to converge toward the ergodic capacity with perfect CSI-TR, which is also plotted in Fig. 5. [[1]]

VII. CONCLUSION

We have studied performance limits of delay-constrained communications with perfect CSI-TR in the low-power regime. Under STPC, we have shown that selection diversity achieves asymptotically the minimum outage performance and provided a simple expression of the outage probability. We have highlighted the gain of CSI-TR over CSI-R only, in terms of outage probability, and shown that this increase
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This page is a continuation of the previous page, but it is not necessary to read the previous page to understand this one. The page contains mathematical equations and explanations related to the capacity and outage probability of communication systems in fading channels.

The text explains that the objective is to verify that the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR, we have characterized the low-power regime. Subject to LTPC, in addition to the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR, we have characterized the low-power regime. Subject to LTPC, in addition to the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR, we have characterized the low-power regime. Subject to LTPC, in addition to the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR, we have characterized the low-power regime. Subject to LTPC, in addition to the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR, we have characterized the low-power regime. Subject to LTPC, in addition to the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR, we have characterized the low-power regime. Subject to LTPC, in addition to the delay-limited capacity and shown that it scales linearly above benefits of fading and CSI-TR.

The present study confirms once again the benefits of communication in the low-power regime. Either in a point-to-point or a multiuser setting, in delay-limited or delay-unlimited communications, the fading along with CSI-TR turn out to be extremely beneficial and provide a substantial gain. Nonetheless, these benefits may be compromised by few technical issues among which the challenging channel estimation and signal decoding in the regime of interest. Circumventing these challenges is certainly not only within our area of expertise, but also depends on advances in microelectronic and signal processing techniques.

**Appendix A**

**Proof of the Asymptotic Equality (22)**

Assume that $F_{\text{max}}(\cdot)$ admits a non null $n$th derivative at $0$, for some $n > 0$. Let $n_0$ be the smallest integer such that $F_{\text{max}}(n_0)(0) \neq 0$. Performing Taylor series expansion of $F_{\text{max}}(\frac{2R}{P})$ at the vicinity of $R = 0$ to the $n_0$th order yields:

$$F_{\text{max}}\left(\frac{2R}{P}\right) = \left(\frac{2}{P}\right)^{n_0} F_{\text{max}}(0) \frac{R^{n_0}}{n_0!} + o_1(R^{n_0}),$$

where $o_1(R^{n_0})$ is a function of $R$ such that $\lim_{R \to 0} o_1(R^{n_0}) = 0$. Our objective now is to verify that $F_{\text{max}}\left(\frac{e^{2MR} - 1}{MP}\right)$ can also be written as

$$F_{\text{max}}\left(\frac{e^{2MR} - 1}{MP}\right) = \left(\frac{2}{P}\right)^{n_0} F_{\text{max}}(0) \frac{R^{n_0}}{n_0!} + o_2(R^{n_0}),$$

where $o_2(R^{n_0})$ is a function of $R$ such that $\lim_{R \to 0} o_2(R^{n_0}) = 0$. For this purpose, we note that

$$F_{\text{max}}\left(\frac{e^{2MR} - 1}{MP}\right)^{(n_0)} = \left(\frac{2}{P}\right)^{n_0} (e^{2MR} - 1)^{(n_0)} \left(\frac{1}{MP}\right)^{(n_0-1)}.$$
APPENDIX B

PROOF OF THEOREM 2

We designate by $\mathcal{L}\{f\}$ the Laplace transform of the function $f$. Let $\phi_m = \mathcal{L}\{f_m\}$, $m = 1, \ldots, M$. Since $n_m$ be the smallest integer such that $F_m^{(n_m)}(0) \neq 0$, then there exist $\eta_m > 0$, such that for any $x \in (0, \eta_m)$, we have:

$$F_m(x) = \frac{F_m^{(n_m)}}{n_m!} x^{n_m} + o(x^{n_m}), \quad (94)$$

$m = 1, \ldots, M$. Thus, for any $x \in (0, \min_{m=1}^{M} \eta_m)$, we have:

$$\sum_{m=1}^{M} F_m(x) \approx \sum_{m=1}^{M} \frac{F_m^{(n_m)}}{n_m!} x^{n_m}. \quad (95)$$

Hence, $P_{out}^{\text{CSI-TR}}(R, P, M)$ is asymptotically equal to:

$$P_{out}^{\text{CSI-TR}}(R, P, M) \approx \sum_{m=1}^{M} \frac{F_m^{(n_m)}}{n_m!} x^{n_m} \left( \frac{2 M R}{P} \right)^{n_m}. \quad (97)$$

where (97) is a consequence of (95) and the fact that $P_{out}^{\text{CSI-TR}}(R, P, M) \to 0$, then we necessarily have $\frac{R}{P} \to 0$. Before computing $P_{out}^{\text{CSI-TR}}(R, P, M)$, we note first that since $F_m^{(n_m)}(0) \neq 0$, then we have:

$$0 \neq F_m^{(n_m)}(0) \quad (98)$$

$$= \lim_{x \to 0} F_m^{(n_m)}(x) \quad (99)$$

$$= \lim_{x \to 0} s F_m^{(n_m)}(s) \quad (100)$$

where (100) follows from the Initial Value Theorem; (101) follows because the PDF of $\sum_{m=1}^{M} \eta_m$ is given by $f_{\sum}(x) = (f_1 * \ldots * f_M)(x)$, where $*$ designates the convolution operator; and (108) holds true since both limits in (103) and (107) exist. Thus, the limit in (108) exists and is non-null (since $F_m^{(n_m)}(0) \neq 0$). But, $\lim_{x \to 0} \frac{F_m^{(n_m)}(s)}{s M_{n_m}} \phi_m(s) \neq 0$ due to (103). Hence $\lim_{x \to 0} \frac{F_m^{(n_m)}(0)}{s M_{n_m}} \neq 0$ which implies that $n = \sum_{m=1}^{M} n_m$. Consequently, $F_m^{(n_m)}(0)$ can be further simplified as:

$$F_m^{(n_m)}(0) = \lim_{x \to 0} \frac{F_m^{(n_m)}}{n_m!} x^{n_m} \cdot \left( \frac{2 M R}{P} \right)^{n_m}. \quad (109)$$

Finally, using (97) and (111), we have:

$$P_{out}^{\text{CSI-TR}}(R, P, M) \approx \frac{1}{M^{\sum_{m=1}^{M} n_m}} \frac{\left( \sum_{m=1}^{M} n_m \right)!}{\prod_{m=1}^{M} n_m!}, \quad (122)$$

which we wanted to prove.

APPENDIX C

PROOF OF LEMMA 2

First, we note that $P(s)$ defined in (10) is strictly increasing on $[0, \infty)$. Indeed, let $s_1$ and $s_2$ be two nonnegative real numbers, such that $s_1 > s_2$. Then we have:

$$P(s_1) - P(s_2) = \int_{s_2}^{s_1} \left[ \sum_{\gamma \in \mathcal{R}(s_1) \cap \mathcal{R}(s_2)} \mathcal{F}_{\mathcal{R}(s)}(\gamma) \right] dF(\gamma) \quad (113)$$

$$> \int_{s_2}^{s_1} \mathcal{F}_{\mathcal{R}(s)}(\gamma) dF(\gamma) \quad (114)$$

$$= s_2 \operatorname{Prob}\{\gamma \in \mathcal{R}(s_1) \cap \mathcal{R}(s_2)\} \quad (115)$$

$$\geq 0. \quad (116)$$

as we claimed. In the above derivation, (113) follows because $\mathcal{R}(s_2) \subset \mathcal{R}(s_1)$ since $s_1 > s_2$; (114) holds true since $\gamma \in \mathcal{R}(s_2)$ and thus $s_2 < \frac{1}{M^{\sum_{m=1}^{M} n_m}} \frac{\left( \sum_{m=1}^{M} n_m \right)!}{\prod_{m=1}^{M} n_m!}$ and also because $\mathcal{R}(s_1) \cap \mathcal{R}(s_2)$ is a set of positive measure (in the sense of Lebesgue).

Second, for $s = 0$, $\mathcal{R}(0) = \emptyset$, the empty set, and thus $P(0) = 0$. Finally, we verify the continuity of $P(s)$ on $[0, \infty)$ as follows. First, we write $P(s)$ as follows:

$$P(s) = \int_{s_2}^{s_1} \left[ \frac{1}{M^{\sum_{m=1}^{M} n_m}} \mathcal{F}_{\mathcal{R}(s)}(\gamma) \right] dF(\gamma), \quad (117)$$

where $\mathcal{F}_{\mathcal{R}(s)}(\gamma)$ is the indicator function of $\mathcal{R}(s)$. Then, we notice that the function $g(\cdot, \cdot)$ defined on $\mathbb{R}^2 \times (0, \infty)$ by:

$$g(\gamma, s) = \frac{1}{s} \left[ \frac{1}{M^{\sum_{m=1}^{M} n_m}} \mathcal{F}_{\mathcal{R}(s)}(\gamma) f(\gamma) \right]$$

is continuous in $s$ for every $\gamma$. Furthermore,

$$|g(\gamma, s)| \leq \frac{1}{s} s \mathcal{F}_{\mathcal{R}(s)} f(\gamma) \quad (118)$$
\[ s = \chi_{[0,\infty)}(\gamma) \]  
\[ f(\gamma) \leq s = \int_{0}^{\infty} s \chi_{[0,\infty)}(\gamma) dF(\gamma) \]

By taking the limits on both sides of (122), we obtain that
\[ \lim_{s \to 0} \mathcal{P}(s) = 0, \]
which is equal to \( \mathcal{P}(0) \). That is, \( \mathcal{P}(s) \) is also continuous at 0.

In summary, we have verified that \( \mathcal{P}(s) \) is a strictly increasing and continuous function on \([0, \infty)\), with \( \mathcal{P}(0) = 0 \). To complete the proof, it suffices to recall that \( s^* (P) \) satisfies (10) and distinguish two cases:

**Case 1:** \( P < \lim_{s \to 0} \mathcal{P}(s) = E \left[ \frac{1}{M} \sum_{m=1}^{M} P_m^* (\gamma) \right] \)

In this case, \( s^* (P) \) is strictly increasing along with the fact that \( \mathcal{P}(0) = 0 \), we obtain
\[ \lim_{s \to 0} s^* (P) = 0. \]

**Case 2:** \( P \geq E \left[ \frac{1}{M} \sum_{m=1}^{M} P_m^* (\gamma) \right] \)

In this case, \( s^* = \sup \{ s : \mathcal{P}(s) < P \} = +\infty \), since \( \mathcal{P}(s) \) is continuous and increasing on \([0, \infty)\).

This completes the proof of Lemma 2.

**References**


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