Analysis of Hybrid Selection/Maximal-Ratio Combining in Rayleigh Fading

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Abstract—In this paper, we use a novel virtual branch technique to succinctly derive the mean and variance of the combiner output signal-to-noise ratio for hybrid selection/maximal-ratio combining in a multipath-fading environment.

Index Terms—Diversity methods, fading channels, maximal-ratio combining, selection diversity, virtual branch technique.

I. INTRODUCTION

HYBRID selection/maximal-ratio combining (H-S/MRC) is a reduced-complexity diversity combining scheme, where \( L \) (with the largest signal-to-noise ratio (SNR) at each instant) out of \( N \) diversity branches are selected and combined using MRC [1]–[6]. This technique provides improved performance over \( L \) branch MRC when additional diversity is available, without requiring additional electronics and/or power consumption.

In this paper, we derive the mean and variance of the combiner output SNR of H-S/MRC for any \( L \) and \( N \) under the assumption of independent Rayleigh fading on each diversity branch with equal SNR averaged over the fading.1 This is performed using a novel “virtual branch” technique, which simplifies the derivation of the mean (derived previously [1]) and permits derivation of the variance.2

II. DIVERSITY COMBINING ANALYSIS

A. Virtual Branch Technique: The Key Idea

The analysis of H-S/MRC based on a chosen ordering of the branches at first appears to be complicated, since the SNR statistics of the ordered branches are not independent. Even the average combiner output SNR calculation alone can require a lengthy derivation, as seen in [1]. Here, we alleviate this problem by transforming the ordered-branch variables into a new set of independently, identically distributed (i.i.d.) virtual branches. This reduces the derivation of the moments of the combiner output SNR to the calculation of the moments of the linear combination of i.i.d. random variables.

1 We assume that instantaneous channel estimation using a scanning receiver across all possible diversity branches is feasible, such as with slow fading. However, H-S/MRC also offers improvement in fast fading conditions, and our results serve as an upper bound on the performance when perfect channel estimates are not available.

2 We extend this study to analyze the symbol error probability performance in [2]–[6].

In this framework, the average SNR of the combined output is obtained in a more concise manner than the derivation given in [1]. Furthermore, the extension to the derivation of the combiner output SNR variance can be made using this virtual branch technique. Note that the mean and variance of the combiner output SNR for the selection diversity (SD) and MRC are special cases of our results.

B. General Theory

Let \( \gamma_i \) denote the instantaneous SNR of the \( i \)th diversity branch defined by

\[
\gamma_i = \frac{\alpha_i E_s}{N_0} \tag{1}
\]

where \( E_s \) is the average symbol energy, and \( \alpha_i \) is the instantaneous gain, and \( N_0 \) is the noise power spectral density of the \( i \)th branch. We model the \( \gamma_i \)'s as continuous random variables with probability density function (pdf) \( f_{\gamma_i}(x) \) and mean \( \Gamma_i = \mathbb{E}\{\gamma_i\} \).

Let us first consider a general diversity-combining (GDC) system with the instantaneous output SNR of the form

\[
\gamma_{\text{GDC}} = \sum_{i=1}^{N} a_i \gamma(i) \tag{2}
\]

where \( a_i \in \{0,1\}, \gamma(i) \) is the ordered \( \gamma_i \), i.e., \( \gamma(1) > \gamma(2) > \cdots > \gamma(N) \), and \( N \) is the number of available diversity branches. It will be apparent later that several diversity combining schemes, including H-S/MRC, turn out to be special cases of (2).

We assume that the instantaneous branch SNR’s \( \gamma_i \) are independent with the same average SNR, i.e., \( \Gamma_i = \Gamma \) for \( i = 1, \cdots, N \). Denoting \( \gamma(\mathbb{N}) = \gamma(1)^T \gamma(2)^T \cdots \gamma(N)^T \), where the superscript \( T \) denotes transpose, the joint pdf of \( \gamma(1), \gamma(2), \cdots, \gamma(N) \) for a Rayleigh fading channel can be derived using the theory of “order statistics” [7] as

\[
f_{\gamma(\mathbb{N})}(\{\gamma(i)\}_{i=1}^{N}) = \left\{ \begin{array}{ll}
\frac{N!}{\Gamma^N} e^{-(1/\Gamma) \sum_{i=1}^{N} \gamma(i)}, & \gamma(1) > \gamma(2) > \cdots > \gamma(N) > 0 \\
0, & \text{otherwise.} \end{array} \right. \tag{3}
\]

The mean SNR of combiner output signal is given by (4), shown at the bottom of the next page.

Since the statistics of the ordered branches are no longer independent, the evaluation of the mean SNR involves nested integrals, which are in general cumbersome and complicated to compute. This can be alleviated by transforming the instantaneous SNR’s of the ordered diversity branches into a
new set of virtual branch instantaneous SNR’s, \( V_n \), using the following relation:

\[
\gamma(i) = \sum_{n=1}^{N} \frac{1}{n} V_n. \tag{5}
\]

The Jacobian \( J \) of the above transformation is \( \Gamma^{N}/N! \). Let \( \gamma(N+1) = 0 \). Then the recursion for \( \gamma(i) \) is

\[
\gamma(i) = \gamma(i+1) + \frac{1}{i} V_i, \quad i = 1, \ldots, N. \tag{6}
\]

Hence, the \( V_n \)'s are not necessarily ordered, and \( 0 < V_n < \infty \).

Denoting \( V_N \triangleq [V_1, V_2, \ldots, V_N]^T \), the joint pdf of \( V_1, V_2, \ldots, V_N \) can be obtained, using the distribution theory for transformations of random vectors [7], as

\[
f_{V_N}(\{v_n\}_{n=1}^{N}) = J f_{\gamma(N)}(\{\gamma(i)\}_{i=0}^{N}) f_{\gamma(i)}(\gamma(i)), \tag{7}
\]

By substituting (5) into (3) along with \( J \), it can be verified after some algebra that (7) becomes

\[
f_{V_N}(\{v_n\}_{n=1}^{N}) = \prod_{n=1}^{N} f_{V_n}(v_n) \tag{8}
\]

where \( f_{V_n}(\cdot) \) is the pdf of \( V_n \) given by

\[
f_{V_n}(v) = \begin{cases} e^{-v}, & 0 < v < \infty \\ 0, & \text{otherwise.} \end{cases} \tag{9}
\]

Therefore, the instantaneous SNR’s of the virtual branches are i.i.d. normalized exponential random variables.\(^3\)

The instantaneous SNR of the combiner output can now be expressed in terms of the instantaneous SNR of the virtual branches as

\[
\gamma_{GDC} = \sum_{n=1}^{N} b_n V_n \tag{10}
\]

where \( b_n = \left( \Gamma/n \right) \sum_{i=1}^{n} a_i \). Using the fact that normalized exponential random variables have unity mean \( \mathbb{E}\{V_n\} = 1 \), the mean of the combiner output SNR can now be calculated as

\[
\Gamma_{GDC} = \mathbb{E}\left\{ \sum_{n=1}^{N} b_n V_n \right\} = \sum_{n=1}^{N} b_n. \tag{11}
\]

Similarly, the variance of the combiner output SNR is

\[
\sigma^2_{GDC} = \text{Var}\left\{ \sum_{n=1}^{N} b_n V_n \right\} = \sum_{n=1}^{N} b_n^2 \tag{12}
\]

where the unit variance of the normalized exponential random variable \( \mathbb{E}\{V_n\} = 1 \) is used. Note that the independence of the virtual branch variables plays a key role in simplifying the derivation of (11) and (12).

\(^3\)When the \( a_i \)'s are not all equal, then the virtual branches are only conditionally independent. This is investigated in [2].

The average SNR gain of the diversity combining compared to a single branch system is a commonly accepted performance measure [8]. This quantity is calculated as

\[
G_{GDC} \triangleq 10 \log_{10} \left\{ \frac{\Gamma_{GDC}}{\Gamma} \right\} = 10 \log_{10} \left\{ \frac{\sum_{n=1}^{N} b_n}{\Gamma} \right\}. \tag{13}
\]

To assess the effectiveness of diversity combining in the presence of multipath, we also define the normalized standard deviation of the combiner output SNR as

\[
\sigma_{GDC} \triangleq 10 \log_{10} \left\{ \frac{\sqrt{\sigma_{GDC}^2}}{\Gamma_{GDC}} \right\} = 10 \log_{10} \left\{ \frac{\sum_{n=1}^{N} b_n^2}{\Gamma_{GDC}} \right\}. \tag{14}
\]

C. H-S/MRC Analysis

In this section, the general theory derived in Section II-B is used to evaluate the performance of H-S/MRC. The instantaneous output SNR of H-S/MRC is

\[
\gamma_{H-S/MRC} = \sum_{i=1}^{L} \gamma(i) \tag{15}
\]

where \( 1 \leq L \leq N \). Note that \( \gamma_{H-S/MRC} = \gamma_{GDC} \) with \( a_i = 1 \) for \( i = 1, \ldots, L \) and \( a_i = 0 \) for \( i = L + 1, \ldots, N \). In this case, \( b_n = \Gamma \) for \( n = 1, \ldots, L \) and \( b_n = \Gamma(L/n) \) for \( n = L + 1, \ldots, N \).

Substituting this into (11) and (12), the mean and the variance of the combiner output SNR can be easily obtained as

\[
\Gamma_{H-S/MRC} = L \left( 1 + \sum_{n=L+1}^{N} \frac{1}{n} \right) \Gamma \tag{16}
\]

and

\[
\sigma^2_{H-S/MRC} = L \left( 1 + L \sum_{n=L+1}^{N} \frac{1}{n^2} \right) \Gamma^2 \tag{17}
\]

respectively. Therefore, the average SNR gain of H-S/MRC in decibels is

\[
G_{H-S/MRC} = 10 \log_{10} \left\{ L \left( 1 + \sum_{n=L+1}^{N} \frac{1}{n} \right) \right\}. \tag{18}
\]
and normalized standard deviation of the combiner output SNR in decibels is

\[
\sigma_{H-S/MRC} = 10 \log_{10} \left\{ \frac{\left[ 1 + L \sum_{n=L+1}^{N} \frac{1}{n^2} \right]}{\sqrt{L \left[ 1 + \sum_{n=L+1}^{N} \frac{1}{n} \right]}} \right\}. \tag{19}
\]

Note that (16) agrees with the result of [1]. Also note that SD and MRC are limiting cases of H-S/MRC with \(L = 1\) and \(L = N\), respectively, and the result of (16) agrees with the well-known results for SD and MRC given by [8, pp. 313–319, eqs. (5.2.8), (5.2.16)].

### III. Numerical Examples

In this section, the results derived in the previous section for H-S/MRC are illustrated. The notation H-\(L/N\) is used to denote H-S/MRC that selects and combines \(L\) out of \(N\) branches. Note that H-1/1 is a single branch receiver, H-1/N is SD, and H-N/N is MRC with \(N\) branches.

Fig. 1 shows the average SNR gain with H-\(L/N\) as a function of \(L\) for various \(N\). The data points denoted by the “squares” represent the gain using H-1/N. The data points denoted by the “stars” represent the gain using H-L/L and serve as a lower bound for the gain of H-L/N. It can be seen that H-L/N can provide a significant gain over H-L/L, when additional diversity is available. Fig. 2 shows the average SNR gain using H-L/N versus \(N\) for various \(L\). The data points denoted by the “squares” and “stars” represent the average SNR gain using H-1/N and H-N/N, respectively. Note that the curves for H-1/N and H-N/N serve, respectively, as a lower and upper bound for the average SNR gain of H-L/N.

Fig. 3 shows the normalized standard deviation of the H-L/N combiner output SNR as a function of \(L\) for various \(N\). The solid curves are parameterized by different \(N\), starting from the highest curve with \(N = 2\), and decrease monotonically to the lowest curve with \(N = 20\).

Fig. 4 shows the normalized standard deviation of the H-L/N combiner output SNR versus \(N\) for various \(L\). It can be clearly
seen that the curves for H-1/N and H-N/N upper and lower bound, respectively, the normalized standard deviation of the H-L/N combiner output SNR.

IV. CONCLUSIONS

We derived the mean and variance of the combiner output SNR of H-S/MRC in a multipath fading environment for any L and N. We analyzed this system using a novel “virtual branch” technique, which resulted in a simple derivation of the mean (derived previously [1]) and permitted derivation of the variance.

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