# Performance of a Non-Orthogonal STBC 

# over Correlative Fading Channels 

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#### Abstract

It has been recently shown that, for non-orthogonal space-time block code (STBC), the multiple-input multipleoutput (MIMO) maximum-likelihood (ML) metric can also be decoupled into single-input single-output (SISO) ML metrics for decoding simplification just as for orthogonal STBC. In this work, we utilized the decoupled metrics of a non-orthogonal STBC to derive the symbol error rate (SER) in correlative fading channels and show that, when the non-orthogonal code is generated by converting an orthogonal code using proper precoding, the conversion will improve the SER performance when the MIMO channels are correlated.


Keywords - space-time block code, MIMO, maximumlikelihood decoding, corrrelative fading channels

## I. INTRODUCTION

DUE to the orthogonal structure of the orthogonal STBC, the decoding of data symbols can be made simplified by decoupling the MIMO ML metric into separate SISO ML metrics [1]-[3]. However, recently, it is found that nonorthogonal STBC can also be decoded by exactly the same simple manner [4]. This means decoupled ML decoding is not the prerogative of the orthogonal STBC. Both nonorthogonal and orthogonal STBCs belong to the class of linear STBCs [2], [4] and new codes for both are yet to be discovered, though more attention has so far been paid to the design of orthogonal STBCs [1], [5]-[7]. A non-orthogonal STBC may be derived from an orthogonal STBC by proper precoding to achieve minimum SER for correlative fading communications [8]. Since minimum error rate is the ultimate performance measure for data communications, we are motivated to investigate the SER performance of nonorthogonal STBCs so formed for correlative fading channels.
In this work, we shall derive symbol error rate (SER) expression for a non-orthogonal STBC over correlative fading channel models. This non-orthogonal code has been
given in [8] and falls into a class of linear STBC specified in [4]. Comparison will be made between the SER performance for an orthogonal STBC and the corresponding nonorthogonal STBC derived from that orthogonal STBC. We will assume QAM signaling over Rayleigh fading channels in our SER derivations.
The rest of the paper is organized as follows: Section II briefly reviews the class of linear STBC introduced in [4]. Section III derives the SER of a non-orthogonal STBC for $M$-QAM transmission over correlative fading channels. Then, Section IV presents numerical examples including performance comparisons between orthogonal and nonorthogonal codes. Finally, Section V draws conclusions.

## II. Non-Orthogonal STBC

We consider a wireless communication system with $P$ transmit antennas and $Q$ receive antennas employing linear STBC transmission. Let the equivalent baseband path gain from the $p$ th transmit antenna to $q$ th receive antenna be $h_{p, q}$, $p=1,2, \ldots, P, q=1,2, \ldots, Q$.

A linear STBC transmission can be described by a $P \times N$ code matrix as [2]

$$
\mathbf{G}=\left[\begin{array}{cccc}
g_{11} & g_{12} & \cdots & g_{1 N}  \tag{1}\\
g_{21} & g_{22} & \cdots & g_{2 N} \\
\vdots & \vdots & \cdots & \vdots \\
g_{P 1} & g_{P 2} & \cdots & g_{P N}
\end{array}\right] .
$$

Here, $g_{p n}$ is the codeword transmitted from the $p$ th transmit antenna at the $n$th time slot, $n=1,2, \ldots, N$, with $N$ time slots constituting a block. Each codeword $g_{p n}$ is a linear
combination of information symbols $\left\{x_{k}\right\}$ and their conjugates $\left\{x_{k}^{*}\right\}, k=1,2, \ldots, K, K \leq N$. In other words, $K$ information symbols over a block of $N$ time slots are chosen for transmission through $P$ transmit antennas. Thus, the code rate is $K / N$. For this work, since we will consider $M$-ary QAM transmission, the $K$ symbols of $\left\{x_{k}\right\}$ used for a block are selected from $M$ possible constellation points. Different block may select different $K$ symbols from the constellation. The code matrix $\mathbf{G}$ can also be expressed alternatively by another two $P \times N$ code matrices $\mathbf{A}_{k}$ and $\mathbf{B}_{k}$ as [2]

$$
\begin{equation*}
\mathbf{G}=\sum_{k=1}^{K}\left(x_{k c} \mathbf{A}_{k}+j x_{k s} \mathbf{B}_{k}\right) \tag{2}
\end{equation*}
$$

where $x_{k c}=\operatorname{Re}\left[x_{k}\right]$ and $x_{k s}=\operatorname{Im}\left[x_{k}\right]$, with $\operatorname{Re}[\cdot]$ and $\operatorname{Im}[\cdot]$ respectively denoting the real and imaginary part.

The received signal vector $\mathbf{r}_{q}=\left[r_{1, q}, r_{2, q}, \cdots, r_{N, q}\right]^{T}, T$ denoting transposition, at the $q$ th receive antenna over the block of $N$ time slots is given by

$$
\begin{equation*}
\mathbf{r}_{q}=\mathbf{y}_{q}+\mathbf{n}_{q}=\mathbf{G}^{T} \mathbf{h}_{q}+\mathbf{n}_{q} \tag{3}
\end{equation*}
$$

where $\mathbf{y}_{q}=\left[y_{1, q}, y_{2, q}, \cdots, y_{N, q}\right]^{T}$ is the noise-free received signal vector, $\mathbf{n}_{q}=\left[n_{1, q}, n_{2, q}, \cdots, n_{N, q}\right]^{T}$ is the zero-mean additive white Gaussian noise (AWGN) vector with covariance matrix $\sigma_{n}^{2} \mathbf{I}_{N}$, with $\mathbf{I}_{N}$ being the $N \times N$ identity matrix , and $\mathbf{h}_{q}=\left[h_{1, q}, h_{2, q}, \cdots, h_{P, q}\right]^{T}$ is the channel gain vector for the $q$ th receive antenna. For (3), we shall assume that $\mathbf{h}_{q}$ remains constant over one block of $N$ time slots (quasi-static fading).
The class of the linear STBC to be discussed here as given in [4] will possess the following properties: for $k=1,2, \ldots, K, i=1,2, \ldots, K$, and $i \neq k$

$$
\begin{align*}
\mathbf{A}_{k} \mathbf{A}_{i}^{H} & =-\mathbf{A}_{i} \mathbf{A}_{k}^{H}  \tag{4a}\\
\mathbf{B}_{k} \mathbf{B}_{i}^{H} & =-\mathbf{B}_{i} \mathbf{B}_{k}^{H}  \tag{4b}\\
\mathbf{B}_{i} \mathbf{A}_{k}^{H} & =\mathbf{A}_{k} \mathbf{B}_{i}^{H} \tag{4c}
\end{align*}
$$

where $H$ denotes Hermitian transposition. The code class given by (4) is more general than the amicable orthogonal design (AOD) defined in [2]. Some non-orthogonal codes that do fall into the class given by (4) have appeared recently in the literature [8], [9] and we shall later use one of these non-orthogonal codes for SER performance demonstrations.
In [4], it is shown that a linear STBC satisfying (4) can be decoded by separable ML metrics (decoupled from the general ML metric) for each complex data symbol. Moreover, the individual metric can further be decomposed into two quadrature metrics, one for the real part of the data
symbol and the other for the imaginary part. For square $M$ QAM signaling, this decomposition would further reduce the computation complexity from order $\mathrm{O}(M)$ to $\mathrm{O}(\sqrt{M})$. The two quadature metrics are given as

$$
\begin{align*}
D_{M L, k c} & =\left(R_{k, 1}^{\prime}-\lambda_{k, 1} x_{k c}^{\prime}\right)^{2}  \tag{5a}\\
D_{M L, k s} & =\left(R_{k, 2}^{\prime}-\lambda_{k, 2} x_{k s}^{\prime}\right)^{2} \tag{5b}
\end{align*}
$$

The mathematical symbols in (5) are illustrated below.
The $k$ th complex data symbol is $x_{k}=x_{k c}+j x_{k s}$. Define

$$
\begin{align*}
& a_{k}^{2}=\sum_{q=1}^{Q}\left\|\mathbf{A}_{k}^{T} \mathbf{h}_{q}\right\|^{2}=\sum_{q=1}^{Q}\left\|\mathbf{a}_{k, q}\right\|^{2}  \tag{6a}\\
& b_{k}^{2}=\sum_{q=1}^{Q}\left\|\mathbf{B}_{k}^{T} \mathbf{h}_{q}\right\|^{2}=\sum_{q=1}^{Q}\left\|\mathbf{b}_{k, q}\right\|^{2}  \tag{6b}\\
& 2 c_{k}=j \sum_{q=1}^{Q} \mathbf{h}_{q}^{H}\left(\mathbf{B}_{k} \mathbf{A}_{k}^{H}-\mathbf{A}_{k} \mathbf{B}_{k}^{H}\right)^{*} \mathbf{h}_{q}  \tag{6c}\\
& R_{k, 1}=\frac{1}{2} \sum_{q=1}^{Q}\left[\mathbf{r}_{q}^{H} \mathbf{A}_{k}^{T} \mathbf{h}_{q}+\left(\mathbf{r}_{q}^{H} \mathbf{A}_{k}^{T} \mathbf{h}_{q}\right)^{H}\right]  \tag{6d}\\
& R_{k, 2}=\frac{j}{2} \sum_{q=1}^{Q}\left[\mathbf{r}_{q}^{H} \mathbf{B}_{k}^{T} \mathbf{h}_{q}-\left(\mathbf{r}_{q}^{H} \mathbf{B}_{k}^{T} \mathbf{h}_{q}\right)^{H}\right] \tag{6e}
\end{align*}
$$

The first three quantities are all real positive [4]. Let $\lambda_{k, 1}$ and $\lambda_{k, 2}$ be the two eigenvalues of the matrix

$$
\left[\begin{array}{cc}
a_{k}^{2} & c_{k}  \tag{7}\\
c_{k} & b_{k}^{2}
\end{array}\right]=\mathbf{U}_{k}\left[\begin{array}{cc}
\lambda_{k, 1} & 0 \\
0 & \lambda_{k, 2}
\end{array}\right] \mathbf{U}_{k}^{H}
$$

where $\mathbf{U}_{k}$ is the $2 \times 2$ unitary matrix whose columns consist of orthonormal eigenvectors respectively corresponding to $\lambda_{k, 1}, \lambda_{k, 2}$. The set of quadrature components of the data symbol and the set ( $R_{k, 1}, R_{k, 2}$ ) are then transformed into new sets by

$$
\begin{align*}
& {\left[\begin{array}{l}
x_{k c}^{\prime} \\
x_{k s}^{\prime}
\end{array}\right]=\mathbf{U}_{k}^{H}\left[\begin{array}{l}
x_{k c} \\
x_{k s}
\end{array}\right],}  \tag{8a}\\
& {\left[\begin{array}{l}
R_{k, 1}^{\prime} \\
R_{k, 2}^{\prime}
\end{array}\right]=\mathbf{U}_{k}^{H}\left[\begin{array}{l}
R_{k, 1} \\
R_{k, 2}
\end{array}\right] .} \tag{8b}
\end{align*}
$$

## III. Ser Analysis for a Non-Orthogonal StBC

We will take a non-orthogonal STBC given in [8] using $M$ QAM signaling over correlative fading channels. The code given in [8] satisfies all the conditions of (4) but with some relaxations, viz., $\mathbf{A}_{k} \mathbf{A}_{k}^{H}=\mathbf{B}_{k} \mathbf{B}_{k}^{H}$ (This does not imply $\mathbf{A}_{k}=\mathbf{B}_{k}$ ) and $\mathbf{B}_{k} \mathbf{A}_{k}^{H}-\mathbf{A}_{k} \mathbf{B}_{k}^{H}=\mathbf{0}$. With the relaxations, we easily find from (6) that $c_{k}=0$,
$\lambda_{k, 1}=a_{k}^{2}=b_{k}^{2}$ (this does not imply $\mathbf{a}_{k, q}=\mathbf{b}_{k, q}$ either), and $\mathbf{U}_{k}=\mathbf{I}_{2}$. Then (5) becomes

$$
\begin{align*}
& D_{M L, k c}=\left(R_{k, 1}-\lambda_{k} x_{k c}\right)^{2}  \tag{9a}\\
& D_{M L, k s}=\left(R_{k, 2}-\lambda_{k} x_{k s}\right)^{2} \tag{9b}
\end{align*}
$$

where $\lambda_{k}=\lambda_{k, 1}=\lambda_{k, 2}$. In light of (9a) and (9b), we can picture the system as formed by equivalent channels with virtual channel gains $\lambda_{k}$ and virtual noisy received signals $R_{k, 1}, R_{k, 2}$ respectively for real and imaginary signal components $x_{k c}, x_{k s}$. Then, (9a) and (9b) are nothing more than just the minimum distance metrics. Thus we can respectively regard $\quad n_{k c}=R_{k, 1}-\lambda_{k} x_{k c} \quad$ and $n_{k s}=R_{k, 2}-\lambda_{k} x_{k s}$ as the real and imaginary parts of a virtual complex noise. In [4], it is shown that

$$
\begin{align*}
& n_{k c}=\frac{1}{2} \sum_{q=1}^{Q}\left[\mathbf{n}_{q}^{H} \mathbf{A}_{k}^{T} \mathbf{h}_{q}+\left(\mathbf{n}_{q}^{H} \mathbf{A}_{k}^{T} \mathbf{h}_{q}\right)^{H}\right]  \tag{10a}\\
& n_{k s}=\frac{j}{2} \sum_{q=1}^{Q}\left[\mathbf{n}_{q}^{H} \mathbf{B}_{k}^{T} \mathbf{h}_{q}-\left(\mathbf{n}_{q}^{H} \mathbf{B}_{k}^{T} \mathbf{h}_{q}\right)^{H}\right] \tag{10b}
\end{align*}
$$

and these two noise components are identical Gaussian RV's with zero mean and variance $\sigma_{k c}^{2}=\sigma_{k s}^{2}=a_{k}^{2} \sigma_{n}^{2} / 2$.

In (6a) and (6b), we can consider $\mathbf{A}_{k}^{T} \mathbf{h}_{q}=\mathbf{a}_{k, q}=\left[a_{k, 1, q} a_{k, 2, q} \cdots a_{k, P, q}\right]^{T} \quad$ and $\quad \mathbf{B}_{k}^{T} \mathbf{h}_{q}=$ $\mathbf{b}_{k, q}=\left[\begin{array}{ll}b_{k, 1, q} & b_{k, 2, q} \cdots b_{k, P, q}\end{array}\right]^{T}$ as the equivalent virtual channel vectors for the $q$ th receive antenna as seen by $x_{k c}$ and $x_{k s}$ respectively. If the original physical channels $\left\{h_{p, q}\right\}$ have complex Gaussian gains, then $\left\{a_{k, p, q}\right\}$ and $\left\{b_{k, p, q}\right\}$ will be correlated complex Gaussian gains as both $\mathbf{a}_{k, q}$ and $\mathbf{b}_{k, q}$ are linear transformations of $\mathbf{h}_{q}$.
For square $M$-QAM signaling over fading channels, $x_{k c}$ and $x_{k s}$ each randomly takes the value from the set $\left\{\left(2 m-1-M_{c}\right) d, m=1,2, \ldots, \sqrt{M} \quad\right\}$ with equal probability $1 / \sqrt{M}$, where $d$ is a constant that can be used for power control. The average transmitted energies of $x_{k c}$ and $x_{k s}$ are given by $E_{c, a v}=E_{s, a v}=\frac{M-1}{3} d^{2}=\frac{d^{2}}{2 g}$, where $g=\frac{3}{2(M-1)}$. Using (9), the received signal energies are $a_{k}^{4} E_{c, a v}=b_{k}^{4} E_{s, a v}$. Therefore, the total
received signal-to-noise ratios (SNRs) after the ML linear processor are given by

$$
\begin{align*}
\gamma_{k c} & =a_{k}^{2} \frac{E_{c, a v}}{\sigma_{n}^{2} / 2}=\sum_{q=1}^{Q} \sum_{p=1}^{P}\left|a_{k, p, q}\right|^{2} \frac{E_{c, a v}}{\sigma_{n}^{2} / 2} \\
& =\frac{2 E_{c, a v}}{\sigma_{n}^{2}} \sum_{l=1}^{L}\left|a_{k, l}\right|^{2}=\sum_{l=1}^{L} \gamma_{k c, l}  \tag{11a}\\
\gamma_{k s} & =\gamma_{k c}=\sum_{l=1}^{L} \gamma_{k s, l}=\sum_{l=1}^{L} \gamma_{k c, l}=\gamma_{k} \tag{11b}
\end{align*}
$$

where we have stacked $a_{k, p, q}$ and $b_{k, p, q}$ respectively into $a_{k, l}$ and $b_{k, l}$ by letting $L=P Q$, and defined $\gamma_{k c, l}=\frac{2 E_{c, a v}}{\sigma_{n}^{2}}\left|a_{k, l}\right|^{2}=\gamma_{k s, l}=\frac{2 E_{s, a v}}{\sigma_{n}^{2}}\left|b_{k, l}\right|^{2}$ as the quadrature branch SNRs. This is equivalent to redefining $\mathbf{a}_{k}=\left[a_{k, 1}, a_{k, 2}, \cdots, a_{k, L}\right]^{T} \quad$ and $\quad \mathbf{b}_{k}=$ $\left[b_{k, 1}, b_{k, 2}, \cdots, b_{k, L}\right]^{T}$ as the new virtual channel vectors respectively seen by $x_{k c}$ and $x_{k s}$.
For fixed set of $\left\{\gamma_{k c, l}\right\}$ and $\left\{\gamma_{k s, l}\right\}$, the conditional SER for square $M$-QAM can be calculated using the moment generating function (MGF)-based approach as [10]
$P_{M}\left(e \mid \gamma_{k}\right)=1-\left[1-P_{\sqrt{M}}\left(e \mid \gamma_{k}\right)\right]\left[1-P_{\sqrt{M}}\left(e \mid \gamma_{k}\right)\right]$
$=P_{\sqrt{M}}\left(e \mid \gamma_{k}\right)+P_{\sqrt{M}}\left(e \mid \gamma_{k}\right)-P_{\sqrt{M}}\left(e \mid \gamma_{k}\right) P_{\sqrt{M}}\left(e \mid \gamma_{k}\right)$
$=\frac{4(\sqrt{M}-1)}{\sqrt{M}} Q\left(\sqrt{2 g \gamma_{k}}\right)-\frac{4(\sqrt{M}-1)^{2}}{M} Q^{2}\left(\sqrt{2 g \gamma_{k}}\right)$
$=\frac{4(\sqrt{M}-1)}{\pi \sqrt{M}} \int_{0}^{\pi / 2} \exp \left(-\frac{g \gamma_{k}}{\sin ^{2} \theta}\right) d \theta$
$-\frac{4(\sqrt{M}-1)^{2}}{\pi M} \int_{0}^{\pi / 4} \exp \left(-\frac{g \gamma_{k}}{\sin ^{2} \theta}\right) d \theta$,
where we have utilized the following identities [10] $Q(x)=\frac{1}{\pi} \int_{0}^{\pi / 2} \exp \left(-\frac{x^{2}}{2 \sin ^{2} \theta}\right) d \theta$
and $Q^{2}(x)=\frac{1}{\pi} \int_{0}^{\pi / 4} \exp \left(-\frac{x^{2}}{2 \sin ^{2} \theta}\right) d \theta$.
The overall average SER for the $k$ th data symbol is obtained by averaging (12) over $\gamma_{k}$ as

$$
\begin{aligned}
P_{M, k} & =\int_{0}^{\infty} P_{M}\left(e \mid \gamma_{k}\right) p\left(\gamma_{k}\right) d \gamma_{k} \\
& =\frac{4(\sqrt{M}-1)}{\pi \sqrt{M}} \int_{0}^{\pi / 2} M_{\gamma_{k}}\left(-\frac{g}{\sin ^{2} \theta}\right) d \theta
\end{aligned}
$$

$$
\begin{equation*}
-\frac{4(\sqrt{M}-1)^{2}}{\pi M} \int_{0}^{\pi / 4} M_{\gamma_{k}}\left(-\frac{g}{\sin ^{2} \theta}\right) d \theta \tag{13}
\end{equation*}
$$

where $M_{\gamma}(s)=\int_{0}^{\infty} p(\gamma) e^{s \gamma} d \gamma$ is the MGF and we have used $\gamma_{k}=\sum_{l=1}^{L} \gamma_{k, l}=\sum_{l=1}^{L} \gamma_{k c, l}=\sum_{l=1}^{L} \gamma_{k s, l}$.

We now need to decorrelate the channels $\mathbf{a}_{k}=\left[a_{k, 1}, a_{k, 2}, \cdots, a_{k, L}\right]^{T}$ and $\mathbf{b}_{k}=\left[b_{k, 1}, b_{k, 2}, \cdots, b_{k, L}\right]^{T}$ to facilitate further computation. The covariance matrices of $\mathbf{a}_{k}$ and $\mathbf{b}_{k}$ can be unitary diagonalized so that a new pair of channel vectors $\mathbf{a}_{k}^{\prime}=\mathbf{U}_{k c}^{H} \mathbf{a}_{k}=\left[a_{k, 1}^{\prime}, a_{k, 2}^{\prime}, \cdots, a_{k, L}^{\prime}\right]^{T}$ and $\mathbf{b}_{k}^{\prime}=\mathbf{U}_{k s}^{H} \mathbf{b}_{k}=\left[b_{k, 1}^{\prime}, b_{k, 2}^{\prime}, \cdots, b_{k, L}^{\prime}\right]^{T}$ can be obtained, where $\mathbf{U}_{k c}$ and $\mathbf{U}_{k s}$ are respectively the unitary matrices that diagonalize the covariance matrices of $\mathbf{a}_{k}$ and $\mathbf{b}_{k}[15]$. Now the components of $\mathbf{a}_{k}^{\prime}$ and $\mathbf{b}_{k}^{\prime}$, viz., $\left\{a_{k, l}^{\prime}\right\}$ and $\left\{b_{k, l}^{\prime}\right\}$ are uncorrelated. As a result, the new equivalent branch SNRs $\left\{\gamma_{k, l}^{\prime}=\gamma_{k c, l}^{\prime}=\gamma_{k s, l}^{\prime}=2\left|a_{k, l}^{\prime}\right|^{2} E_{c, a v} / \sigma_{n}^{2}\right\}$ are uncorrelated [11] and independent if the original physical channels $\left\{h_{p, q}\right\}$ are Gaussian. Thus, the MGF of the combined SNRs $\gamma_{k}^{\prime}=\sum_{l=1}^{L} \gamma_{k, l}^{\prime}$ is equal to the products of MGFs of the branch SNRs. We get

$$
\begin{equation*}
M_{\gamma_{k}^{\prime}}\left(-\frac{g}{\sin ^{2} \theta}\right)=\prod_{l=1}^{L} M_{\gamma_{k, l}^{\prime}}\left(-\frac{g}{\sin ^{2} \theta}\right) \tag{14}
\end{equation*}
$$

As shown in [11], $\sum_{l=1}^{L} \gamma_{k, l}=\gamma_{k}=\gamma_{k}^{\prime}=\sum_{l=1}^{L} \gamma_{k, l}^{\prime}$. Thus, $M_{\gamma_{k}^{\prime}}\left(-\frac{g}{\sin ^{2} \theta}\right)=M_{\gamma_{k}}\left(-\frac{g}{\sin ^{2} \theta}\right)$. Using this along with (14), we can rewrite (13) as

$$
\begin{align*}
P_{M, k} & =\frac{4(\sqrt{M}-1)}{\pi \sqrt{M}} \int_{0}^{\pi / 2} \prod_{l=1}^{L} M_{\gamma_{k, l}^{\prime}}\left(-\frac{g}{\sin ^{2} \theta}\right) d \theta \\
& -\frac{4(\sqrt{M}-1)^{2}}{\pi M} \int_{0}^{\pi / 4} \prod_{l=1}^{L} M_{\gamma_{k, l}^{\prime}}\left(-\frac{g}{\sin ^{2} \theta}\right) d \theta \tag{15}
\end{align*}
$$

If the original physical channels $\left\{h_{p, q}=h_{p, q, c}+j h_{p, q, s}\right\}$ are zero mean Gaussian RVs, so will be the virtual channels seen by $x_{k c}$ and $x_{k s}$, i.e., $\left\{a_{k, l}^{\prime}=a_{k c, l}^{\prime}+j a_{k s, l}^{\prime}\right\}$ and $\left\{b_{k, l}^{\prime}=b_{k c, l}^{\prime}+j b_{k s, l}^{\prime}\right\}$. Then both $x_{k c}$ and $x_{k s}$ will experience Rayleigh fading. The MGF is given by [11]

$$
\begin{equation*}
M_{\gamma_{k}^{\prime}}(s)=\frac{1}{\prod_{l=1}^{L}\left(1-\bar{\gamma}_{k, l}^{\prime} s\right)} \tag{16}
\end{equation*}
$$

where $\bar{\gamma}_{k, l}^{\prime}=E\left(\gamma_{k, l}^{\prime}\right)$ is the average SNR of the $l$ th virtual branch with $E(\cdot)$ denoting expectation.

It is important to note that, since $\mathbf{A}_{k}$ and $\mathbf{B}_{k}$ are different for different $k$, so are $a_{k}$ and $b_{k}$, and hence, different data symbols may experience different SER $P_{M, k}$.
We now briefly describe the generation of the nonorthogonal code as given in [8] used for the above SER derivation. The non-orthogonal code that fits the description here can be derived from an orthogonal STBC with code matrix $\mathbf{G}$ by precoding as

$$
\begin{equation*}
\mathbf{C}=\left(\mathbf{D}_{f} \mathbf{U}_{h}^{H}\right)^{T} \mathbf{G} \tag{17}
\end{equation*}
$$

where $\mathbf{U}_{h}$ is the unitary matrix that diagonalizes the channel covariance matrix $\mathbf{R}_{h h}=E\left[\mathbf{h h}^{H}\right]$ (for Rayleigh fading, $E[\mathbf{h}]=\overline{\mathbf{h}}=\mathbf{0}$ ), with $\mathbf{h}=\left[h_{1}, h_{2}, \ldots, h_{L}\right]$ being the $L \times 1$ vector formed by stacking $\left\{h_{p, q}\right\}, L=P Q$, and $\mathbf{D}_{f}$ (a function of SNR and $\mathbf{R}_{h h}$ ) is a diagonal matrix whose diagonal elements are obtained by optimum power loading [8, eq. (30)]. For a given $\mathbf{U}_{h}$ and $\mathbf{D}_{f}$, it can be readily verified that $\mathbf{C C}^{H}$ (see example given in the next section) is no more diagonal. Thus a non-orthogonal code is generated and this non-orthogonal code will have $\mathbf{A}_{k} \mathbf{A}_{k}^{H}=\mathbf{B}_{k} \mathbf{B}_{k}^{H}$ and $\mathbf{B}_{k} \mathbf{A}_{k}^{H}-\mathbf{A}_{k} \mathbf{B}_{k}^{H}=\mathbf{0}$.

## IV. A Numerical Example

Our example uses a non-orthogonal STBC that is obtained by precoding a rate $3 / 4$ COD code in [5] given by

$$
\mathbf{G}=\left[\begin{array}{cccc}
-x_{3} & 0 & x_{1}^{*} & x_{2}^{*}  \tag{18}\\
-x_{2} & x_{1}^{*} & 0 & -x_{3}^{*} \\
x_{1} & x_{2}^{*} & x_{3}^{*} & 0
\end{array}\right]
$$

For demonstration, we shall use three transmit antennas and a single receive antenna and the constant channel correlation model [12]. Constant correlation can be obtained by a circularly symmetric three-element antenna array with close spacing between elements [12]. The channel covariance matrix is then given by

$$
\mathbf{R}_{h h}=\left[\begin{array}{ccc}
\sigma_{h_{1}}^{2} & \sigma_{h_{1}} \sigma_{h_{2}} \rho & \sigma_{h_{1}} \sigma_{h_{3}} \rho  \tag{19}\\
\sigma_{h_{1}} \sigma_{h_{2}} \rho & \sigma_{h_{2}}^{2} & \sigma_{h_{2}} \sigma_{h_{3}} \rho \\
\sigma_{h_{1}} \sigma_{h_{3}} \rho & \sigma_{h_{2}} \sigma_{h_{3}} \rho & \sigma_{h_{3}}^{2}
\end{array}\right],
$$

where $\sigma_{h_{l}}^{2}=E\left[\left|h_{l}\right|^{2}\right], l=1,2,3$ and $\rho$ is the constant correlation coefficient between any pair of the three channels. In our simulations, we select three values for this coefficient, $\rho=0.3,0.6,0.9$ and the ratios between channel variances as $\sigma_{h_{1}}^{2}: \sigma_{h_{2}}^{2}: \sigma_{h_{3}}^{2}=1: 1: 0.6$. For the case of $\rho=0.6$, the unitary matrix $\mathbf{U}_{h}$ can be readily calculated as

$$
\mathbf{U}_{h}=\left[\begin{array}{ccc}
-0.6335 & 0.7071 & -0.3141  \tag{20}\\
-0.6335 & -0.7071 & -0.3141 \\
-0.4442 & 0 & 0.8959
\end{array}\right]
$$

For a 20 dB average received SNR per channel and $\rho=0.6$, the $\mathbf{D}_{f}$ can be computed as

$$
\mathbf{D}_{f}=\left[\begin{array}{ccc}
0.6935 & 0 & 0  \tag{21}\\
0 & 0.5561 & 0 \\
0 & 0 & 0.4580
\end{array}\right]
$$

while the average received SNR per channel is defined as $\frac{1}{L} \sum_{l=1}^{L} E\left(\left|h_{l}\right|^{2}\right) E\left(\left|x_{k}\right|^{2}\right) / \sigma_{n}^{2}$. Using (18), (20), and (21), we can verify that $\mathbf{C} \mathbf{C}^{H}=\left(\mathbf{D}_{f} \mathbf{U}_{h}^{H}\right)^{T} \mathbf{G} \mathbf{G}^{H}\left(\mathbf{D}_{f} \mathbf{U}_{h}^{H}\right)^{*}=$ $\sum_{k=1}^{K}\left|x_{k}\right|^{2} \mathbf{U}_{h}^{*} \mathbf{D}_{f}^{2} \mathbf{U}_{h}^{T}$ is no more diagonal. The SER vs. average received SNR performance for square 16-QAM over correlative Rayleigh fading channels using the above data is presented in Fig. 1. Theoretical curves using (15) and Monte Carlo simulated curves are both given. It is seen that theoretical results are in excellent agreement with Monte Carlo simulated results. Note that small correlation yields better performance resulting from larger diversity. Then, Fig. 2 presents the comparisons between the SER curves for the orthogonal code of (18) and the corresponding nonorthogonal code for $\rho=0.6$ and $\rho=0.9$. Clearly, we see that the non-orthogonal code outperforms the orthogonal counterpart.

## V. Conclusion

We derive the SER for a non-orthogonal STBC coupled with QAM signaling in correlative fading channels. The nonorthogonal code is formed by precoding an orthogonal STBC using optimum power loading. MIMO ML metric decoupling into SISO metrics is still applicable to the non-orthogonal code. Simulation results of SER performances for a code example in excellent agreements with theoretical results show that non-orthogonal codes outperform the orthogonal counterparts when MIMO channels are correlated.


Figure 1. SER performance for a rate $3 / 4$ non-orthogonal STBC using square 16-QAM over Rayleigh fading with constant channel correlations. Monte Carlo simulations are marked by circles.


Figure 2. SER performance comparisons between the rate $3 / 4$ orthogonal and corresponding non-orthogonal STBC over Raleigh fading with constant channel correlations.

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