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Achievable Rates of ZF Receivers in Massive MIMO with Phase Noise Impairments

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Abstract—The effect of oscillator phase noise on the sum-rate performance of large multi-user multiple-input multiple-output (MU-MIMO) systems is studied. A Rayleigh fading MU-MIMO uplink channel is considered, where channel state information (CSI) is acquired via training. The base station (BS), which is equipped with an excess of antenna elements, $M$, uses the channel estimate to perform zero-forcing (ZF) detection. A lower bound on the sum-rate performance is derived. It is shown that the proposed receiver structure exhibits an $O(\sqrt{M})$ array power gain. Additionally, the proposed receiver is compared with earlier studies that employ maximum ratio combining and it is shown that it can provide significant sum-rate performance gains at the medium and high signal-to-noise-ratio (SNR) regime. Further, the expression of the achievable sum rate provides new insights on the effect of various parameters on the overall system performance.

I. INTRODUCTION

In the last decades there has been an unprecedented increase in demand for high data rates over wireless cellular systems. Multiple-input multiple-output (MIMO) systems have attracted significant attention since they promise substantial gains in the capacity of wireless networks [1], [2]. Such systems have gained popularity and are already part of modern cellular communication standards. The demand for increased high speed wireless access is expected to accelerate in the years to come. In addition, power consumption at the base stations (BS) of cellular systems is a major problem. Therefore, future cellular systems are required to exhibit increased spectral efficiency as well as improved energy efficiency.

Recently, it has been shown that using an excess of BS antenna elements, $M$, (Massive MIMO) can boost data rates and simultaneously it can reduce the required transmit power with very simple linear signal processing [3], [4]. In [5] the authors show that for the case of MU-MIMO uplink channel with linear receivers, one can reduce the total transmitted power proportionally to $M$ (or $\sqrt{M}$) for the case of perfect (or imperfect) channel state information (CSI), without compromising the total throughput. Further studies also highlight the gains in energy efficiency that are offered by Massive MIMO [6], [7].

In practice, communication systems inevitably suffer from transceiver impairments. One such impairment is phase noise. Phase noise is introduced by imperfect oscillators that convert the baseband signal to the passband and vice versa. Ideal oscillators output a sinusoid that is stable in amplitude, frequency and phase. However, due to imperfections of the circuitry of the oscillators the phase of the sinusoid drifts randomly. Consequently, phase noise can significantly degrade the performance of coherent communication. In a system where CSI is acquired via uplink training, phase noise causes a partial loss of coherency between the estimated channel coefficients and the channel realizations during detection. The large array gain offered by Massive MIMO is a consequence of the coherent combining of received signals using the estimated channel gains. Therefore, the study of the effect of phase noise impairment on the performance of Massive MIMO systems is particularly important.

In earlier work [8], we have studied the effect of oscillator phase noise in the frequency selective Massive MIMO uplink when detection is performed via time-reversal maximum-ratio-combining (TR-MRC). TR-MRC is optimal at low signal-to-noise-ratio (SNR), where the system performance is limited by thermal noise. However, in the high-SNR regime the system performance is interference limited and the MRC sum-rate performance saturates. Zero-forcing (ZF) receivers can provide improved performance in this regime since they effectively suppress multi-user interference. In this work we study the sum-rate performance of ZF receivers with phase noise impairments and compare it with the sum-rate performance of MRC receivers. To the best of authors’ knowledge, this paper is the first to report such a study.

We consider a Rayleigh fading channel with $K$ non-cooperative users and $M$ BS antennas. Phase noise is introduced both at the transmitter and receiver side and is modelled as a Wiener phase noise process. The BS estimates the channel via uplink training and uses the acquired CSI to perform ZF detection. We provide an analytical formula for a lower bound on the sum-capacity and a coding strategy that achieves this lower bound. We prove that, just like MRC, ZF also offers an $O(\sqrt{M})$ array power gain. We show that ZF is superior to MRC in term of sum-rate performance in the high-SNR regime. For a desired per-user information rate, we also study the extra per-user transmit power required by MRC receivers when compared to the ZF receiver. The derived achievable sum-rate can provide new insights in the design of Massive MIMO systems with phase noise impairments.

II. SYSTEM MODEL

We consider a frequency flat multi-user multiple-input multiple-output (MU-MIMO) uplink channel, with $K$ single antenna users and $M$ base station (BS) antenna elements. The channel between user $k$ and antenna element $m$ is given by $\sqrt{d_k} h_{m,k}$, where $d_k$ and $h_{m,k}$ models the slow and fast fading components, respectively. In this work, we assume

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a block fading model, where $h_{m,k}$ is fixed for a block of $K + N_D$ channel uses and varies independently from one block to another. Since $d_k$ changes at a much slower rate compared to $h_{m,k}$, we assume that it is fixed and known throughout the communication. The fast fading coefficients $h_{m,k}$ are assumed to be independent identically distributed (i.i.d.) circularly symmetric complex normal $\mathcal{CN}(0,1)$ random variables.

Phase noise is introduced at the transmitter during up-conversion, when the baseband signal is multiplied with the carrier generated by the oscillator. The phase of the generated carrier drifts randomly, resulting in phase distortion of the transmitted signal. A similar phenomenon also happens at the receiver side during down-conversion of the bandpass signal transmitted. A similar phenomenon also happens at the receiver side during down-conversion of the bandpass signal generated by the oscillator. The phase of the generated carrier drifts randomly, resulting in phase distortion of the transmitted signal. A similar phenomenon also happens at the receiver side during down-conversion of the bandpass signal transmitted.

Based uplink transmission scheme. In the proposed scheme, we assume that the phase noise processes at the BS antennas at time $n$ are independent identically distributed zero-mean Gaussian increments. $f_c$ is the carrier frequency, $T_s$ is the symbol interval and $c$ is a constant that depends on the oscillator. Similarly, we can define the phase noise processes at the $K$ users.

Let $x_k[i]$ be the symbol transmitted from the $k$-th user at time $i$. The received signal at the $m$-th BS antenna element at time $i$ is then given by

$$y_m[i] = \sqrt{P} \sum_{k=1}^{K} \sqrt{d_k} h_{m,k} e^{-j(\phi[i] - \theta_k[i])} x_k[i] + n_m[i],$$

where $n_m[i] \sim \mathcal{CN}(0, \sigma_n^2)$ is additive white Gaussian noise (AWGN). Each user transmits a stream of i.i.d. $\mathcal{CN}(0, 1)$ information symbols (i.e. $x_k[i] \sim \mathcal{CN}(0, 1)$), that are independent of the information symbols of the other users. $P$ denotes the average uplink transmitted power from each user.

### III. TRANSMISSION SCHEME AND ZF RECEIVER

Motivated by the need for low-complexity channel estimation and detection algorithms, we propose the following block based uplink transmission scheme. In the proposed scheme, a transmission block of $K + N_D$ channel uses consists of $K$ channel uses (for uplink channel estimation) followed by the data phase (for data transmission) of duration $N_D$ channel uses. For coherent demodulation, the BS needs to estimate the uplink channel. This is facilitated through the transmission of uplink pilot symbols during the training phase of each transmission block. The users transmit uplink training signals sequentially in time, i.e. at any given time only one user is transmitting uplink training signals and all other users are idle. To be precise, the $k$-th user sends an impulse signal of amplitude $\sqrt{P_k} K$ at the $(k-1)$-th channel use and is idle for the remaining portion of the training phase. Here, $P_k$ is the average power transmitted by a user during the training phase. Therefore, using (1) the signal received at the $m$-th BS receiver at time $k - 1$, $k = 1, \ldots, K$ is given by

$$y_m[k - 1] = \sqrt{P_k} K d_k h_{m,k} e^{-j(\phi[k-1] - \theta_k[k-1])} + n_m[k - 1].$$

#### A. LMMSE Channel Estimation

Based on the received signal during training, the effective MIMO channel is estimated in order to enable coherent detection of the received symbols. The received symbols during training, given by (2), can be expressed in the following matrix-vector form

$$Y = \sqrt{P_k} K H \Omega D^{1/2} + N$$

where $[Y]_{m,k} = y_m[k - 1], [H]_{m,k} = h_{m,k}, \Omega = \text{diag}(e^{-j(\phi[0]-\theta[0])}, \ldots, e^{-j(\phi[K-1]-\theta[K-1])})$, $D = \text{diag}(d_1, \ldots, d_K)$, $d_k > 0$, $\forall k = 1, \ldots, K$ and $[N]_{m,k} = n_m[k - 1]$. Also, observe that $\Omega \Omega^H = \Omega^H \Omega = I_K$. We define the effective MIMO channel matrix as $G \triangleq \Omega H$. The LMMSE estimate $\hat{G}$ for $G$ based on (3) is given by [11]

$$\hat{G} = Y \text{diag} \left\{ \frac{\sqrt{P_k} K d_1}{P_k K d_1 + \sigma^2}, \ldots, \frac{\sqrt{P_k} K d_K}{P_k K d_K + \sigma^2} \right\}.$$  

As expected, the channel estimate is distorted by the AWGN and by the phase noise at the transmitter and at the BS. In the following, we treat this channel estimate as the true channel and perform equalization based on that. Conditioned on $\Omega$, it is clear from (3) that the entries of $Y$ are Gaussian distributed. Since $G = H \Omega$ and $\hat{G}$ is a column wise scaled version of $Y$ (see (4)), it follows that conditioned on $\Omega$, $G$, $\hat{G}$ and $\hat{E} \triangleq \hat{G} - G$ are all Gaussian distributed. Therefore, conditioned on $\Omega$, the LMMSE estimate in (4) is actually the MMSE estimate and therefore $\hat{G}$ and $\hat{E}$ are independent (conditioned on $\Omega$). Since we are interested in evaluating the performance of ZF receivers, the equalizer would simply be the pseudo-inverse of $\hat{G}$, i.e. $\hat{\Omega} \triangleq \left( \hat{G}^H \hat{G} \right)^{-1} \hat{G}^H$. Since $\hat{G}$ and $\hat{E}$ are independent (conditioned on $\Omega$), it therefore follows that

$$\mathbb{E}_{G,\hat{E} \mid \Omega} [\hat{G}^H \hat{E}] = \mathbb{E}_{G \mid \Omega} [\hat{G}^H] \mathbb{E}_{E \mid \Omega} [E] = 0$$

since $\mathbb{E}_{E \mid \Omega} [E] = 0$.

#### B. Zero-Forcing (ZF) Equalization

During the data interval, the received symbol at time $i \geq K$ at the $m$-th BS antenna is given by

$$y_m[i] = \sqrt{P_D} \sum_{k=1}^{K} h_{m,k} e^{-j(\phi[i] - \theta_k[i])} \sqrt{d_k} x_k[i] + n_m[i].$$

---

1The discrete-time phase noise model is used since we are interested in the discrete-time complex baseband representation of the transmit and receive signals.
Therefore, the matrix-vector form of the channel input-output relation is given by
\[
y[i] = \sqrt{P_D} H \Phi[i] D^{1/2} x[i] + n[i],
\]
where \( \Phi[i] \triangleq \text{diag} \{e^{-j(\phi[i]-\theta_2[i])}, \ldots, e^{-j(\phi[i]-\theta_K[i])}\} \) and \( P_D \) is the per-user average transmit power constraint during the data interval. In general, the relation between \( P_D \) and \( P_p \) can be arbitrary. However, in this case we restrict to the case where \( P_p = \beta P_D, \beta > 0 \).

Since \( G = H \Omega \) and \( \Omega \Omega^H = I_K \), it follows that \( H = G \Omega^H \). Also, since \( G = G - \mathcal{E} \), we have \( H = G \Omega^H = (G - \mathcal{E}) \Omega^H \). Therefore, replacing \( H \) by \( (G - \mathcal{E}) \Omega^H \) in (7) we get
\[
y[i] = \sqrt{P_D} (G - \mathcal{E}) \Omega^H \Phi[i] D^{1/2} x[i] + n[i].
\]
(8)

Then, the detected symbol vector \( \hat{x}[i] \) at time \( i \) is given by
\[
\hat{x}[i] = G^\dagger y[i] = \sqrt{P_D} \Omega^H \Phi[i] D^{1/2} x[i]
- \sqrt{P_D} e_k^\top G^\dagger \Omega^H \Phi[i] D^{1/2} x[i] + e_k^\top G^\dagger n[i],
\]
(9)
where \( G^\dagger \triangleq (G^\top G)^{-1} G^\top \) is the pseudo-inverse of \( G \).

IV. ACHIEVABLE RATES

In this work we are interested in evaluating the sum-rate performance for the system under study. In the following, we describe the approach to derive an achievable sum rate for the Massive MIMO uplink with phase noise impairments when ZF reception is employed. From (9) the detected symbol for the \( k \)-th user at time \( i \) is given by
\[
\hat{x}_k[i] = \sqrt{P_D} d_k e_j(\phi[k-1]-\theta_k[k-1]) e^{-j(\phi[i]-\theta_k[i])} x_k[i] - \sqrt{P_D} e_k^\top G^\dagger \Omega^H \Phi[i] D^{1/2} x[i] + e_k^\top G^\dagger n[i],
\]
(10)
where \( e_k \) is the \( K \)-dimensional all-zero column vector that has a single 1 at position \( k \). Let \( A_k[i] \triangleq \sqrt{P_D} d_k e_j(\phi[k-1]-\theta_k[k-1]) e^{-j(\phi[i]-\theta_k[i])} \). Since the exact phase noise realizations are unknown, the desired signal symbol \( x_k[i] \) is rotated by an unknown phase. Further, the statistics of \( A_k[i] \) depends on \( i \). However, the statistics of the phase noise processes is known to the BS and as a result it knows the mean value of \( A_k[i] \). In (10) we add and subtract the term \( \mathbb{E}[A_k[i]] x_k[i] \). The added term \( \mathbb{E}[A_k[i]] x_k[i] \) is the desired signal term and the term \((A_k[i] - \mathbb{E}[A_k[i]]) x_k[i] \) is relegated to an effective noise term. Therefore, (10) can be expressed as
\[
\hat{x}_k[i] = \mathbb{E}[A_k[i]] x_k[i] + \mathbb{E}N_k[i] \quad (11)
\]
where
\[
\mathbb{E}N_k[i] \triangleq \mathbb{E}F_k[i] + \text{MUI}_k[i] + \text{AN}_k[i]
\]
\[
F_k[i] \triangleq (A_k[i] - \mathbb{E}[A_k[i]]) x_k[i],
\]
\[
\text{MUI}_k[i] \triangleq -\sqrt{P_D} e_k^\top G^\dagger \Omega^H \Phi[i] D^{1/2} x[i],
\]
\[
\text{AN}_k[i] \triangleq e_k^\top G^\dagger n[i].
\]

Proposition 1: In (11) the desired signal term \( \mathbb{E}[A_k[i]] x_k[i] \) and the effective additive noise term \( \mathbb{E}N_k[i] \) are uncorrelated.

Proof: It is straightforward to see that \( \mathbb{E}[A_k[i]] x_k[i] \mathbb{E}F_k[i] = 0 \). Also \( \mathbb{E}[A_k[i]] x_k[i] \text{MUI}_k[i] = 0 \) by the assumption that \( x[i] \) and \( n[i] \) are uncorrelated. Finally,
\[
\mathbb{E}[MUI_k[i] x_k[i] \mathbb{E}[A_k[i]]] = -\sqrt{P_D} \mathbb{E}[e_k^\top G^\dagger \Omega^H \Phi[i] D^{1/2} e_k]
\]
\[
\cdot \mathbb{E}[A_k[i]] = -\sqrt{P_D} d_k \Omega \left[ e_k^\top \mathcal{G} \Omega \mathcal{E} \right] \mathbb{E}[A_k[i]] e^{-j(\phi[i]-\theta_k[i])}.
\]
which follows since \( \mathbb{E}[\mathcal{G} \Omega \mathcal{E}] \mathcal{G}^\dagger = 0 \) from (5). Hence, we can conclude that \( \mathbb{E}[A_k[i]] x_k[i] \mathbb{E}N_k[i] = 0 \).

We also provide a result that will be proved useful in the derivation of the achievable rates.

Theorem 1: The mean value of \( A_k[i] \) and the variance \( \text{Var}(\mathbb{E}N_k[i]) \triangleq \mathbb{E}[\mathbb{E}N_k[i]^2] - \mathbb{E}[\mathbb{E}N_k[i]]^2 \) are given by
\[
\mathbb{E}[A_k[i]] = \sqrt{P_D} d_k (1 - \kappa_k^2[i]),
\]
\[
\text{Var}(\mathbb{E}N_k[i]) = P_D d_k (1 - \kappa_k^2[i]) + \sigma^2 C_k,
\]
where \( \kappa_k[i] = e^{-\sigma^2 \theta_k(i-k-1)} \) and
\[
C_k \triangleq \left( \sum_{l=1}^K \frac{P_p d_l}{\alpha \beta} K d_l + 1 \right) \frac{\beta^2 P_p K d_k + 1}{(M-K)\beta^2 P_p K d_k}.
\]

In the following, we describe a method to derive an achievable information rate for the \( k \)-th user. Similar techniques have been used earlier in [12], [13], [8]. Observe that, in a given coherence interval, \( \mathbb{E}[A_k[i]] \) and \( \text{Var}(\mathbb{E}N_k[i]) \) are different for different \( i \). However \( \mathbb{E}[A_k[i]] \) and \( \text{Var}(\mathbb{E}N_k[i]) \) for a fixed \( k \) and \( i \) are the same across multiple coherence intervals. Further, the realizations of \( \mathbb{E}N_k[i] \) for a fixed \( k \) and \( i \) are i.i.d. across multiple coherence intervals. Therefore, for the \( k \)-th user, we propose to encode information using \( N_D \) different channel codes, one channel code for each \( i \). That is, in the \( i \)-th channel code information is coded across the \( i \)-th channel use of each coherence interval.

We proceed by describing the derivation of a lower bound for the rate of the \( i \)-th channel code for the \( k \)-th user. Since the BS has knowledge of the statistics of the channel coefficients and the phase noise processes but not of the exact realization, the term \( \mathbb{E}[A_k[i]] \) can be computed and is a constant depending on the user \( k \) and the channel use \( i \) during the data transmission interval. On the other hand, the exact probability distribution of the effective noise term is complicated to compute. However, since the input symbols are uncorrelated to the effective noise \( \mathbb{E}N_k[i] \) (from Proposition 1), a lower bound on the achievable information rate can be derived by considering the worst case uncorrelated noise having the same variance as \( \mathbb{E}N_k[i] \). Given that the input symbols are Gaussian, the worst case uncorrelated additive noise is also zero mean Gaussian with variance equal to the variance of \( \mathbb{E}N_k[i] \). Therefore, the mutual information between
the detected symbol $\hat{x}_k[i]$ and the transmitted symbol is lower bounded by
\[ I(\hat{x}_k[i]; x_k[i]) \geq \log_2 \left( 1 + \frac{(\mathbb{E}[A_k[i]])^2}{\text{Var}(\mathbb{E}[A_k[i]])} \right). \]  

(14)

**Theorem 2:** An achievable rate for the $i$-th channel code of the $k$-th user is given by
\[ R^i_k = \log_2 \left( 1 + \frac{P_0}{\sigma^2} d_k \kappa_k[i]^2 \right). \]

(15)

**Proof:** The result follows directly by substituting (12) and (13) into (14).

The information rate for the no-phase noise case (i.e. $c = 0$, perfect oscillator) can be derived by substituting $\kappa_k[i] = 1$ in (15), and is given by
\[ R_k = \log_2 \left( 1 + \frac{P_0}{\sigma^2} d_k \kappa_k[i]^2 \right). \]

(16)

For a sake of comparison, we also derive the information rate expression for maximum-ratio-combining (MRC) when MMSE channel estimation is used (using similar arguments to the ZF case). The $i$-th code of user $k$ achieves the rate
\[ R_k = \log_2 \left( 1 + \frac{P_0}{\sigma^2} d_k \kappa_k[i]^2 \right). \]

(17)

where $C_k^{\text{MRC}}$ is defined in (16). We conclude the section by defining the average achievable sum-rate for ZF as
\[ R^z_k = \frac{1}{N_D + K} \sum_{k=1}^{K} \sum_{i=K}^{N_D + K - 1} R^i_k, \]

(18)

where the normalization factor $\frac{1}{N_D + K}$ accounts for the loss of spectral efficiency due to training. Similarly, one can derive the average achievable sum-rate for the MRC and no-phase noise cases.

**V. RESULTS - DISCUSSION**

In this section we use the expressions derived in (15), (16), (17) and (18) to present our main results. In the plots we always consider $D = I_K$, which corresponds to the case of equal path loss for all the users, whereas in the theoretical derivations we consider an arbitrary, known and deterministic slow fading matrix $D$. We start with a proposition on the information rate achievable at high SNR.

**Proposition 2:** In the high-SNR regime the rate that user $k$ can achieve at the $i$-th channel use saturates to the value
\[ \lim_{P_0 \to \infty} R^i_k \to \log_2 \left( 1 + \frac{\kappa_k[i]}{1 - \kappa_k[i]} \right). \]

(19)

for ZF and at the value
\[ \lim_{P_0 \to \infty} R^{\text{MRC}}_k \to \log_2 \left( 1 + \frac{d_k \kappa_k[i]}{\frac{1}{M} \sum_{i=1}^{K} d_i} \right) \]

(20)

for MRC.

A desirable property of Massive MIMO is the array power gain they offer, facilitating the design of power efficient communication systems. In earlier work [5], [8] it has been shown that in the case of Massive MIMO with imperfect CSI, one can scale down the total transmit power by $\sqrt{M}$ as the number of BS antennas $M$ increase, while maintaining a fixed positive desired information rate to each user. In this work, we extend the result to the case of Massive MIMO uplink impaired with phase noise when ZF receivers are employed.

**Proposition 3:** An $O(\sqrt{M})$ array power gain is achievable. This implies that a fixed non-zero per user information rate can be achieved if the total transmitted power is reduced by $1.5$ dB and at the same time the number of BS antenna elements is doubled.

**Proof:** From (15) and by the substitution $P_D = E_{\alpha \sigma^2}$, $\alpha > 0$ we have that
\[ R^i_k = \log_2 \left( 1 + \frac{E_{\alpha \sigma^2} d_k \kappa_k[i]}{\frac{1}{M} \sum_{i=1}^{K} d_i} + M^\alpha C_k \right). \]

We look for the largest $\alpha > 0$ such that $\lim_{M \to \infty} R_k[i] \to r > 0 \iff \lim_{M \to \infty} M^\alpha C_k \to \text{const.}$

\[ M^\alpha C_k = M^\alpha \left( \sum_{i=1}^{K} \frac{E_{\alpha \sigma^2}}{\beta \sigma^2} d_i (K d_i + 1) + \frac{\kappa_k[i]}{1 - \kappa_k[i]} \beta K \frac{E_{\alpha \sigma^2}}{\sigma^2} K d_k + \frac{\kappa_k[i]}{1 - \kappa_k[i]} \beta K \frac{E_{\alpha \sigma^2}}{\sigma^2} K d_k \right) + \frac{\kappa_k[i]}{1 - \kappa_k[i]} \beta K \frac{E_{\alpha \sigma^2}}{\sigma^2} K d_k \]

As $M \to \infty$, the first term converges to a finite positive constant if and only $\alpha \leq 1$, whereas the second term converges to a finite positive constant if and only if $\alpha \leq 1/2$. For $\alpha > 1/2$, the second term is unbounded. Therefore, as $M \to \infty$, with $P_D = E_{\alpha \sigma^2}$, we have
\[ \lim_{M \to \infty} R_k[i] \to \begin{cases} \log_2 \left( 1 + \frac{E_{\alpha \sigma^2} d_k \kappa_k[i]}{\frac{1}{M} \sum_{i=1}^{K} d_i} \right), & \alpha = 1/2 \\ \log_2 \left( 1 + \frac{\kappa_k[i]}{1 - \kappa_k[i]} \right), & \alpha < 1/2 \\ 0, & \alpha > 1/2 \end{cases} \]

In Fig. 1 we plot the minimum required $P_0/\sigma^2$ to achieve a fixed desired per-user information rate $r = 1.5$ bpcu as a function of the number of BS antennas $M$ for fixed $K = 10$ users, $N_D = 100$ channel uses, $T_s = 10^{-6}$ s and $c = 4.7 \times 10^{-18}$ (rad Hz)$^{-1}$. We plot the curves for ZF and MRC with
and without phase noise. It is clear that at large $M$, when the number of BS antennas doubles the minimum required power decreases by 1.5 dB. This verifies the $O(\sqrt{M})$ array power gain as stated in Proposition 3.

Further, the superior performance of the ZF over MRC in the high-SNR regime is clear. When it is desired to achieve a per-user rate of $r = 1.5 \text{ bpcu}$ with $\frac{P_D}{\sigma^2} = -10$ dB, ZF with phase noise requires the use of about $M = 160$ BS antennas. However, the MRC system with phase noise for the same parameters requires approximately 80 additional BS antennas. This motivates us to study the $\frac{P_D}{\sigma^2}$ gap between ZF and MRC reception. In Fig. 2 we plot this gap as a function of the desired per-user information rate for fixed $M = 200$ BS antennas and $N_D = 150$ channel uses for various values of the number of users, $K$. It is clear that there is a significant gain in the minimum required power for per-user information rates larger than 1 bpcu. The gain is more pronounced as the number of users increases. The reason for this behaviour is that for $r > 1$ bpcu and for $K$ sufficiently large, MRC is limited by interference, whereas the ZF receiver effectively suppresses multi-user interference yielding better performance. At a per-user rate of $r \approx 0.9$ bpcu there is a turning point. Below this value we operate in the low spectral efficiency regime and the system is limited by the thermal noise. As a result MRC exhibits better performance whereas ZF underutilizes the available degrees of freedom.

In conclusion, we studied the effect of oscillator phase noise in a frequency flat Massive MIMO uplink channel when ZF equalization is employed at the BS. We provided an analytical expression on the achievable sum-rate and we described a coding strategy that achieves the derived sum-rate. Based on the derived expression we proved that an $O(\sqrt{M})$ array power gain is still attainable. Further, we showed that even in the presence of phase noise, ZF equalization performs significantly better than MRC at high spectral efficiencies. The derived achievable sum-rate expressions can provide additional insight on the effect of oscillator phase noise in Massive MIMO systems with ZF equalization.

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