A Novel Multiuser Detection and Channel Estimation Method for Uplink Communication in MIMO-OFDM IDMA System

Dolly Sharma, Rabindra Kumar Singh, Himanshu Katiyar

Abstract: Demand of wireless cellular communication systems has grown rapidly which necessitates high speed and reliable communication for various multimedia applications. Several technologies have been introduced to meet the desired communication requirements such as 1G, 2G, 3G, and 4G but these techniques suffer from multiple-access and interference issues. Hence in this work, we focus on these challenges to improve the cellular communication performance. The main objective of this work is to develop a novel approach for multiuser detection and introducing a new architecture of channel estimation. However, several techniques are present to achieve these objectives but computational complexity and reliable performance still remains an issue. Moreover, interference remains a challenging task. Hence, in order to overcome these issues, we present a joint approach of multiuser detection where we compute the LLRs, find the Gaussian distribution and measure the mean and variance of this distribution. Similarly, in next phase, channel estimation is performed with the help of ESE and DEC. The performance of proposed approach is measured in terms of BER for varied simulation configurations. The experimental analysis reported significant improvement in the performance of MIMO-OFDM IDMA system.

I. INTRODUCTION

During last two decade, we have noticed significant growth in wireless cellular communication and technologies. The next generation wireless communication demand for high speed and reliable communication to support various multimedia applications. However, multiple users communicate in the same environment where it becomes a challenging issue to deliver the reliable communication due to interference and multi-path fading [1]. Several wireless cellular communication standards are present such as 1G, 2G, 3G and 4G [2]. Moreover, now-a-days, LTE (Long Term Evaluation) and LTE-A (Long Term Evaluation-Advanced) techniques [3] have been introduced to satisfy the demand of high data rate communication. Incorporating multiple access (MA) techniques in these communication models is considered as a promising solution to improve the performance. Currently, several multiple access schemes are present such as time-division multiple access (TDMA) [3], Frequency-Division Multiple Access (FDMA) [4], Code-Division Multiple access (CDMA) [5], orthogonal frequency-division multiple access (OFDMA) [6], non-orthogonal frequency-division multiple access (NOMA) [7] and Interleave Division Multiple Access (IDMA) [8]. The IDMA systems are considered as a promising technique to improve the communication performance. The IDMA based models are used in MIMO [9], NOMA [10], and OFDM [11] etc. The IDMA systems have significant impact on 5G communication where huge numbers of users are present. The IDMA takes advantage of its own chip-level interleaver to separate the multiple users at the receiver. The IDMA receiver contains the elementary signal estimator (ESE) which is based on the soft rake processing approach and performs the chip level detection at the receiver end. Once the ESE task is finished, the detected chip LLRs are processed through de-interleaving and de-spreading to obtain the bit level LLRs. Later, the obtained LLRs are given as input to the channel decoder. Moreover, the de-interleaver and data despreading process is considered as a channel decoding process which helps to minimize the error. Currently, IDMA has gained attraction because it can be applied in various communication systems such as multi-carrier transmission, MIMO transmission, joint detection and channel estimation schemes. In [12] IDMA systems are implemented with multiple access and non-Gaussian noise. Bilim et al. [13] reported that the IDMA achieves good performance with flat fading and additive white noise. However, these systems achieve poor performance for multi-path fading scenario because multiple paths reduce the convergence capability of ESE. In order to overcome this issue multiuser detection schemes are developed [11][14-17]. Compared to CDMA, multiuser detection algorithms are easy to implement in IDMA systems. However, the convergence speed of these approaches becomes a tedious task which depends on the signal to interference and noise ratio (SINR). Generally, value of SINR is low during initial iterations but increases slowly, thus, the numbers of iterations required for computation are more in IDMA systems. On other hand, the computational complexity at the receiver end is a challenging issue. According to the conventional architecture of IDM, it has to detect all the signal of other available users for further iterative receiving process at user end. In this context, the channel estimation techniques are implemented [18-20]. The computational complexity for multiple users is a key issue in these types of models. On the other hand, the interleavers are used for distinguishing the signals from different users but it is not convenient for one user to know all parameters of other user’s interleaving parameters.

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Moreover, the IDMA systems suffer from the co-channel and inter-symbol interference which leads to degraded performance for uplink and downlink communication in IDMA system. Currently, cooperative schemes are widely studied in these applications such as Bilim et al. [21] cooperative decode and Forward cooperative scheme using best relay selection. In another study, presented in [22], authors introduced cooperative relay beamforming scheme in IDMA. Similarly, Vishvakse nan et al. [23] presented relay aided cooperative downlink communication approach. In this work, we focus on the uplink communication and developed a novel approach for uplink communication along with channel coding and interleaving. The main contribution of this work is as follows:
(a) Development of an uplink IDMA system
(b) Presenting the chip-by-chip detection model for multi-path uplink communication.
(c) Development of channel coding and interleaving scheme to mitigate the error rate.

Rest of the manuscript is organized in following sub-sections: section II presents brief discussion on current literature in the field of IDMA systems. Section III presents proposed solution to improve the overall performance. Section IV presents experimental and comparative analysis using proposed approach and finally, section V gives concluding remarks.

II. LITERATURE SURVEY

This section presents a brief discussion about recent techniques to improve the performance of IDMA systems. Multiuser detection, beamforming, cooperative communication and relay selection etc. are the significant research components of IDMA. Sharma et al. [11] introduced an approach for multi-user detection and interference cancellation in IDMA. According to this approach spatial log-likelihood multiuser detection is carried out and tested for MIMO-IDMA systems. Moreover, authors have developed modulation sub-carrier classification approximation model. Li et al. [14] discussed that the conventional CBC (Chip-by-Chip) detectors consider that the sensor activity is precisely known. However, this is not convenient for all receivers hence, authors proposed a multiuser detection approach in sporadic IDMA for uplink transmission. In order to achieve this, compressed sensing is applied for MUD activity analysis and data detection. Later, compressed sensing and CBC are combined and CS-CBC detector is developed. Similar to this, Shikula et al. [24] introduced a combined multiuser detection and channel estimation approach for uplink MIMO-OFDM IDMA communication. The main concept of this work is based on the iterative receiver where MUD and channel estimation techniques are performed jointly. This is achieved by computing log-likelihood ratios of coded bits (bit LLRs). During this process, soft decision-directed channel estimation (SDCE) is applied where bit LLRs are used along with the initial channel estimation. Later, tap selection process is presented to improve the channel estimation. Tap selection is used to extract the dominant paths. Several techniques have been reported for channel estimation in OFDM-IDMA systems. TaşPINAR et al. [25] presented Least Squares (LS) and Minimum Mean Square Error (MMSE) based MUD approach to estimate the channel frequency response. Peng et al. [26] discussed that the uplink OFDM-IDMA systems suffer from the carrier frequency offsets. In order to overcome this issue, authors suggested an approach for iterative estimation and cancellation of addition interference. Chen et al. [27] introduced iterative interference cancellation multiuser detection (IIC-MUD) approach for Overloaded MIMO OFDM IDMA System. In [19] Chang et al. presented a combined approach to time tracking and channel estimation for IDMA. In this work, authors have implemented pilot-assisted turbo-type iterative detection model which can be used for recovering the message of each user from the obtained superimposed signals.

On other hand, cooperative schemes are also widely studied in this field. Generally, cooperative schemes are based on the concept of relay selection where two schemes decode –forward (DF) and amplify –forward (AF) are used. Zhou et al. [28] presented superposition modulated cooperative scheme for multiple source node scenario. According to this approach, the source nodes are divided into several cooperative subsets in a Time Division Duplexing (TDD) manner. During transmission, the signal is superimposed which contains its own data information and data obtained from the previous subset node. The node subsets are differentiated using IDMA to receive the signal. Yang et al. [29] presented network coding concept in IDMA and introduced a cooperative relay selection scheme. This work uses Turbo coding model for multi-path systems. Moreover, this approach adopts different interlever for different users and also network coding cooperative schemes are studied. The cooperative approach is used for beamforming in IDMA where optimal relay selection is carried out. Shukla et al. [22] focused on beamforming and relay selection in IDMA. In this work, IDMA and cooperative relay aided beamforming is developed. This approach is divided into two phases where first of all, users broadcast the message, the relay nodes perform the amplification task and forwards the data to the destination node. In order to mitigate the effect of multiple access interference (MAI), the IDMA scheme is incorporated at relay and destination nodes. This can be achieved using efficient beamforming. Hence, authors presented a beam forming approach to minimize the transmit power and controls the QoS parameters. Moreover, the power minimization is not considered as convex optimization problem hence semi-definite problem is formulated and solved using CVX solver.

Similarly, researchers have presented error correcting codes for minimizing the bit error rates and improve the QoS performance. Zhang et al. [30] focused on power and linear density parity check (LDPC) model optimization to improve the IDMA performance. In order to achieve this, the power profiling is considered as main task which are obtained using mutual information between users. Further, a Gaussian approximation method is applied to compute the probability density function. According to the receiver architecture, the detector and decoder exchange the information in an iterative process. Zhang et al. [31] developed raptor-like quasi-cyclic low density parity check (RL-QC-LDPC) to improve the IDMA system for 5G communication applications.
The conventional signal estimator processes data symbol-by-symbol manner whereas in this process, the signal estimator is replaced by a newly designed maximum a posteriori probability multi-user detector.

This section presents a brief discussion about recent techniques of multiple access using IDMA. Several new architectures are discussed along with new interlaver designs to improve the communication performance. Moreover, IDMA schemes are also incorporated with MIMO OFDM systems to satisfy the communication requirement. In addition to this, several researchers have presented forward error correcting techniques to reduce the packet error. However, due to increasing complexity and demand, these systems fail to achieve the desired performance in terms of bit error rate and complexity. Hence, there is a need to present a novel approach to meet the performance requirement.

III. PROPOSED MODEL

This section presents the proposed solution to improve the performance of IDMA systems. In order to achieve this, we divide the proposed solution into two phases as (a) low-complexity multi-user detection and (b) channel estimation. The proposed model provides a joint solution for MUD and channel estimation.

(a) Multiuser detection

Let us consider an IDMA system with K users as depicted in figure 1 which has recursive systematic convolutional (RSC) code model (rate of RSC is $R = \frac{1}{2}$) to protect the $b_k[n]$ information bits of k user. The obtained coded bits are permuted using interlavers $I_k$ and modulated using QPSK modulation scheme. The current baseband signal can be expressed as:

$$r(t) = \sum_{l=0}^{L-1} h_l(t) y(t - \tau_l) e^{j \phi_l(t)} + w(t) \tag{1}$$

where $y(t)$ denotes the transmitted signal expressed as $[t_y(t), \sum_{k=1}^{K} x_k(t)]$ where $t_y$ denotes the Binary Phase Shift Keying (BPSK) modulated training sequence, at this stage signals of all users are combined, $h_l(t)$ denotes the fading coefficient for path $l$ at time $t$, the total path delay is $\tau_l$, total number of path $L$ and phase $\phi$, $w(t)$ denotes the additive Gaussian noise, and $\phi_l(t)$ represents the phase distortion.

$$b_k(n) \ x_k(n) \ c_k(n) \ \ x_k(n)$$

$$\begin{align*}
\text{ENC} & \quad \text{S} & \quad L_k & \quad \text{OQPSK} \\
b_k(n) & \quad c_k(n) & \quad z_k(n) & \quad y(n) \\
\end{align*}$$

Recently, interference cancellation and multiuser detection techniques are adopted to mitigate the effect of interference, MAI and MUD [11]. According to these schemes, detector considers the multipath signal and detects each signal individually by computing log-likelihood ratios (LLRs). These LLRs are aggregated as $(L + 1)$ LLRs for each chip which helps to achieve the more accurate information of signal. After processing through MUD models, the interference of $k$ users $\eta_{k,l}(n)$ can be represented as Gaussian random variable which can be expressed as:

$$\mathbb{F}(n + 1) = h_{k,l}(n + 1) = |h_{k,l}(n)|^2 x_k(n) + \eta_{k,l}(n)$$

$\eta_{k,l}(n)$ denotes the remaining interference of MUD models. From above expression, the outcome of detector can be approximated with the help of Gaussian distribution. Here, we compute the mean and variance of the baseband signal as:

$$E(r(n)) = \sum_{k,l} h_{k,l}(n) E[x_k(n - l)], \forall n \tag{3}$$

To compute the variance, we consider the total number of bits and the data bits in physical propagation path. The difference of bit is computed and mean value is obtained for baseband signal. Similar task is performed in eq.(5) where we compute the mean and variance for interference signal.

Similarly, the covariance of this baseband signal is computed as:

$$\text{Cov}[r(n)] = \sum_{k,l} H_{k,l}(n) \text{Cov}[x_k(n - l)] H_{k,l}^*(n) + \sigma^2 A \mathcal{Y}_n$$

where $\sigma$ is the noise variance and $A$ represents the $2 \times 2$ identity matrix and $H_{k,l}(n) = \left[ h_{k,l}(n) \ -h_{k,l}^*(n) \ h_{k,l}(n) \ h_{k,l}^*(n) \right]$. Similarly, the mean and variance of interference is computed as given in Eq. (5) and (6)

$$\text{Var}[\eta_{k,l}(n)] = H_{k,l}^*(n) [\text{Cov}[r(n + l)] - H_{k,l}(n) \text{Cov}[x_k(n)] H_{k,l}^*(n)] H_{k,l}(n), \forall k,n \tag{7}$$

In the next phase, we generate the LLRs and combine them to detect the multi user signals. This is expressed as:

$$L_{\text{MUD}}[x_k(n)] = 2|h_{k,l}(n)|^2 \frac{\mathcal{F}^T(n + l) - \mathcal{F}^T(n)}{\text{Var}^T(z_k(n))}, \forall k,n$$

$\mathcal{F}^T$ denotes the real part of the received signal. For multiple paths, it can be rewritten as:

$$L_{\text{MUD}}[x_k(n)] = \sum_{l=0}^{L-1} L_{\text{MUD}}[x_k(n)] \tag{8}$$

In order to obtain the accurate information, the priori means and signal variances of the transmitted signal are required which are computed using Eq (6) with the help of extrinsic LLRs of the decoder, given as:

$$E[x_k(n)] = \tan \left( \frac{L_{\text{dec}}[x_k(n)]}{2} \right) + j \tan \left( \frac{L_{\text{dec}}[x_k(n)]}{2} \right)$$

and

$$\text{Cov}[x_k(n)] = \begin{bmatrix} 1 - E[x_k(n)]^2 & 0 \\ 0 & 1 - E[x_k(n)]^2 \end{bmatrix}$$

Here, $x_k^1$ and $x_k^2$ are uncorrelated real and imaginary part of the $x_k(n)$. $E[.]$ and $\text{Var}[.]$ denotes the mean and variance of the signal. This approach can be extended for multiple path to perform the MUD.

(b) Channel Estimation

In this section, we present the proposed solution for channel estimation for IDMA systems. In this stage, we consider that the receiver module has an idle channel state information. Figure 2 shows an architecture of proposed channel estimator.
A Novel Multiuser Detection and Channel Estimation Method for Uplink Communication in MIMO-OFDM IDMA System

In order to estimate the channel, we present pilot embedding model for a synchronous channel without memory. For simplicity, we rewrite Eq. (1) as:

\[ r = \sum_{k=1}^{K} h_k x_k + n \]  

(10)

Let \( P \) be the pilot for \( k \) user denoted as \( p_k \equiv \{ p_k(1), \ldots, p_k(N) \} \), here, we assume that the \( \{ p_k \} \) are generated randomly and remain independent.

After considering the transmitted signal from \( k \)-th user as \( x_k \) and pilot \( p_k \), the received signal can be represented as:

\[ r = \sum_{k=1}^{K} h_k (x_k + p_k) + n \]  

(11)

\( h_k \) is the channel coefficient for \( k \)-th user.

The soft estimate of this signal \( x_k \) is expressed as \( \hat{x}_k \equiv \{ \hat{x}_k(1), \ldots, \hat{x}_k(N) \} \). Hence, the soft estimate \( \hat{h}_k \) can be generated as:

\[ \hat{h}_k = \frac{(x_k + p_k)^H (r - \sum_{k'}=k \hat{h}_{k'} (x_{k'} + p_{k'}))}{(x_k + p_k)^H (x_k + p_k)} \]

The subtraction method denotes the soft-cancellation method and denominator represents the data normalization. In order to improve the system performance, we introduce a novel iterative process where \( \hat{x}_k \) and \( \hat{h}_k \) are initialized from 0 for each user \( k \). With the help of Eq. (12), the receiver model initializes the channel estimating process. According to this process, the ESE and DEC functions are applied by considering \( \hat{h}_k \) as the current channel information. The outcome of these two functions is used for updating the \( \hat{x}_k \) by computing the mean value of the DEC outcome. The signal updating process is defined as:

\[ \hat{x}_{k,j} = \tanh \left( \frac{1}{2} l_{\text{DEC}}(x_{k,j}) \right) \quad \forall k, j \]

(1)

Where \( l_{\text{DEC}} \) is expressed as:

\[ l_{\text{DEC}}(x_{k,j}) \equiv e_{\text{ESE}}(x_{k,j}) + e_{\text{DEC}}(x_{k,j}) \quad \forall k, j \]

\( e_{\text{ESE}}(x_{k,j}) \) is a posteriori information about the signal which is the outcome of DEC function. Later, the channel estimator estimates the coefficients of the channel and iterates until the entire process is finished.

Thus, we obtain the combined model of multiuser detection and channel estimation to improve the performance of IDMA system. The proposed architecture can be incorporated with MIMO-OFDM systems.

IV. RESULTS AND DISCUSSION

In this section, we present the experimental analysis of proposed approach and compared the performance with existing techniques. The complete simulation study is carried out using MATLAB simulation tool running on windows platform with 8GB RAM and i5 64 bit processor. Complete simulation parameters are presented in below given table 1.

<table>
<thead>
<tr>
<th>Parameter name</th>
<th>Considered Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of Subcarrier</td>
<td>64</td>
</tr>
<tr>
<td>Modulation</td>
<td>QPSK</td>
</tr>
<tr>
<td>Fading</td>
<td>Rayleigh</td>
</tr>
<tr>
<td>Path Length</td>
<td>16</td>
</tr>
<tr>
<td>FFT size</td>
<td>128</td>
</tr>
<tr>
<td>Cyclic Prefix</td>
<td>32</td>
</tr>
</tbody>
</table>

The performance of proposed model is measured in terms of BER (Bit Error Rate) for both coded and Uncoded IDMA scenarios along with Gaussian and Non-Gaussian noise considerations. Similarly, we have measured the performance with different interleaving methods such as tree, power and random interleaving methods. Later, we have measured the performance for varied number of users. Figure 3 shows a comparative study in terms of BER vs. SNR where we compare the performance of proposed model with coded IDMA Gaussian, coded IDMA non-Gaussian, Uncoded IDMA Gaussian and Uncoded IDMA non-Gaussian.

![Fig.3. BER vs. SNR comparison for coded, Uncoded IDMA with Gaussian and Non-Gaussian noise.](image)

According to figure 3, we obtain improved performance in terms of BER. The proposed model is based on the coding scheme to minimize the error. This comparative study is adopted from [12]. The obtained BER values are presented in table 1.
Table 1 BER vs. SNR performance for coded and Uncoded IDMA

<table>
<thead>
<tr>
<th></th>
<th>Coded Gaussian</th>
<th>Coded-Non Gaussian</th>
<th>Uncoded Gaussian</th>
<th>Uncoded-Non Gaussian</th>
<th>Proposed IDMA</th>
</tr>
</thead>
<tbody>
<tr>
<td>BER</td>
<td>0.0162057</td>
<td>0.13916</td>
<td>0.110589</td>
<td>0.157289</td>
<td>0.072007</td>
</tr>
<tr>
<td></td>
<td>0.123519</td>
<td>0.097559</td>
<td>0.066127</td>
<td>0.075012</td>
<td>0.021114</td>
</tr>
<tr>
<td></td>
<td>0.083984</td>
<td>0.058618</td>
<td>0.031167</td>
<td>0.025002</td>
<td>0.003528</td>
</tr>
<tr>
<td></td>
<td>0.024386</td>
<td>0.027871</td>
<td>0.010020</td>
<td>0.004777</td>
<td>0.000086</td>
</tr>
<tr>
<td></td>
<td>0.026537</td>
<td>0.060247</td>
<td>0.001895</td>
<td>0.000382</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>0.009163</td>
<td>0.001754</td>
<td>0.000145</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>0.00108</td>
<td>0.000139</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>

This comparative study shows that the average BER performance is obtained as 0.058, 0.041, 0.027, 0.032 and 0.012 using Coded Gaussian, Coded-Non-Gaussian, Uncoded-Gaussian, Uncoded-Non Gaussian and Proposed IDMA, respectively. The proposed model follows iterative process which helps to reduce the bit error rates whereas conventional IDMA systems are based on the CBC detector which fails to achieve the desired performance. The proposed model shows performance improvement by 20.81%, 29.06%, 44.12% and 37.02 when compared with Coded Gaussian Coded-Non-Gaussian, Uncoded-Gaussian, and Uncoded-Non Gaussian, respectively.

Table 2 BER vs. SNR for different Interleaving Methods

<table>
<thead>
<tr>
<th>Method</th>
<th>8-Users</th>
<th>16-User</th>
<th>32-Users</th>
<th>64-User</th>
</tr>
</thead>
<tbody>
<tr>
<td>Tree</td>
<td>0.054303</td>
<td>0.066111</td>
<td>0.072311</td>
<td>0.103008</td>
</tr>
<tr>
<td>Power</td>
<td>0.03158</td>
<td>0.041203</td>
<td>0.046521</td>
<td>0.058803</td>
</tr>
<tr>
<td>Random</td>
<td>0.015849</td>
<td>0.022629</td>
<td>0.026624</td>
<td>0.028808</td>
</tr>
<tr>
<td>Tree</td>
<td>0.006564</td>
<td>0.010541</td>
<td>0.013087</td>
<td>0.01156</td>
</tr>
<tr>
<td>Power</td>
<td>0.002114</td>
<td>0.00396</td>
<td>0.005274</td>
<td>0.003573</td>
</tr>
<tr>
<td>Random</td>
<td>0.00049</td>
<td>0.001123</td>
<td>0.001639</td>
<td>0.000784</td>
</tr>
<tr>
<td>Tree</td>
<td>7.36E-05</td>
<td>0.00022</td>
<td>0.000362</td>
<td>0.00011</td>
</tr>
<tr>
<td>Power</td>
<td>0.002114</td>
<td>0.00396</td>
<td>0.005274</td>
<td>0.003573</td>
</tr>
<tr>
<td>Random</td>
<td>0.00049</td>
<td>0.001123</td>
<td>0.001639</td>
<td>0.000784</td>
</tr>
</tbody>
</table>

With the help of this experiment, we identify that the average BER is obtained as 0.015, 0.020, 0.023, and 0.029 for 8, 16, 32, and 64, respectively.

V. CONCLUSION

In this work, we have focused on mitigating the effect of interference due to multiple access and channel estimation to improve the system performance. In order to achieve this, we consider the IDMA as base model and presented a joint approach for multiuser detection and channel estimation. The multiuser detection scheme is based on the LLR computation where mean and variance of Gaussian distribution is used for MUD. In the next phase, we present the channel estimation process where we apply Rake based soft estimation technique.
Final outcome is obtained with the help of ESE and DEC of the IDMA. The performance of proposed approach is compared with different simulation parameters to prove the robustness of proposed model.

REFERENCES


