Performance Analysis of Low-Density Parity-Check Codes Aided Block Diagonalization using Space Time Block Codes for MIMO Systems

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Abstract: Multiple Input Multiple Output (MIMO) systems are used to assure high spectral efficiency, huge data rate, improved channel capacity and low error rates. Forward error correction coding is required in communication systems for the efficient transmission of data. One of such code is Low-Density Parity-Check codes. Block Diagonalization is used as a precoding technique which reduces Multi-User Interference and Space-Time Block Codes decreases the signal fading effects. The combination of both can effectively mitigate channel impairments. Hence, for the MIMO systems, we have proposed a technique which is Low-Density Parity-Check codes aided Block Diagonalization using Space-Time Block Codes by assuming that the channel state information is known. Low-Density Parity-Check codes provide efficient coding when compared with the other encoding techniques even in rank deficient scenarios.

Index Terms: Binary Phase Shift Keying (BPSK), Block Diagonalization (BD), Inter Symbol Interference (ISI), Low-Density Parity-Check (LDPC) codes, Minimum Mean Square Error (MMSE), Multi-User Interference (MUI), Signal to Noise Ratio (SNR), Quadrature Phase Shift Keying (QPSK), Quadrature Amplitude Modulation (QAM), Singular Value Decomposition (SVD), Space Time Block Codes (STBCs), Trellis Coded Modulation (TCM) and Zero Forcing (ZF)

I. INTRODUCTION

Wireless Communications (in particular mobile communications) are spread extensively due to their enormous usage. MIMO systems with the help of various diversity techniques are used to maintain the expanding data rate by providing high spectral efficiency. MIMO systems are used to ensure tremendous increase in speed of transmission with least possible degradation in Quality of Service [1], [2], [3]. The idea of numerous antennas instead of single antenna is justified to be successful for the enhancement of bit rate, security, coverage and overall performance of communication systems. By using MIMO systems, channel capacity is increased because total antennas used are directly comparable with the channel capacity. Implementing Spatial Diversity scheme in MIMO results in the increase of link reliability and reduces Bit Error Rate (BER) [4]. Some of the challenging issues which are to be considered in designing of MIMO systems are Multi-User Interference and signal fading effects. By taking these challenges into consideration we have proposed this technique. MUI can be reduced by Block Diagonalization technique by making the efficient use of the channel matrix.

LDPC codes are one of the best channel encoding/decoding codes and which are capable of data transmission with greater efficiency. Extensively used in Space Communications due to their channel impairing capabilities [5]. The outstanding performance of these codes is due to the decoding algorithms which consumes huge amount of memories [6]. Block Diagonalization technique is mainly based on SVD technique and Water Filling algorithm [7].

SVD technique decomposes a single matrix to three matrices [8]. This technique is also known as factor analysis. Applying SVD technique to matrix B, it can be decomposed as

\[
B = USV^H
\]  

where, B represents \( p \times q \) matrix, 

\( U \) represents \( p \times q \) left-hand unitary matrix, 

\( S \) represents \( q \times q \) diagonal matrix containing singular values of matrix B, 

\( V \) represents \( q \times q \) right-hand unitary matrix.

Water Filling algorithm [9], [10] helps in raising the power to the required level. It can be used in wireless communication such that we can overcome the transmission losses which are ensued, as the power level of the transmitted signal is raised. Hence, by increasing the power level we can able to retrieve the original signal.

In STBCs we will transmit copies of data to reduce signal fading effects [11], [12]. This is done by transmitting the same information twice in two time slots. Then even if the signal in one time slot is corrupted with more noise at a singular instant of time we may able to retrieve the signal from the second time slot as it may not be effected with that much of noise at that particular instant of time. Hence, by reducing signal fading effects we may able to retrieve the original signal.

At the receiver side, ZF and MMSE detectors are generally used for retrieving the information from the signal received at the receiver antenna. ISI can be removed by a Zero Forcing detector. Its performance is ideal in the...
noiseless channel. But, if it is a noisy channel then it will amplify the noise to a great amount at the frequencies where the channel response has less magnitude. ZF estimates the transmitted bits using [13], [14]

\[ R_{zf} = (H^H H)^{-1} H^H D \]  

(1.2)

where, \( H \) represents channel matrix, 
\[ H = \begin{bmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \end{bmatrix} \] for 2x2 MIMO systems, 
\( H^H \) represents Hermitian of the channel matrix, 
\( D \) represents received bits after the addition of noise at the receiver side, 
\( R_{zf} \) represents estimated transmitted bits by ZF method.

As ZF equalizer provides noise amplification for lesser channel response, a more balanced equalizer used in this case is Minimum Square Error equalizer. Unlike ZF equalizer it doesn’t eliminate ISI. But, the total power of the noise and ISI components present in the output are minimized. MMSE retrieves bits by using [15]

\[ R_{mmse} = \frac{1}{\sigma^2} (H^H H + \sigma^2 I)^{-1} H^H D \]  

(1.3)

where, \( H \) represents Channel Matrix, 
\[ H = \begin{bmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \end{bmatrix} \] for 2x2 MIMO systems, 
\( H^H \) represents Hermitian matrix of the channel matrix, 
\( \sigma^2 \) represents variance, 
\( I \) represents Identity Matrix, 
\( D \) represents received signal matrix, 
\( R_{mmse} \) represents estimated transmitted bits by MMSE method

II. SYSTEM MODEL

The block diagram shown in fig 2.1 represents the transmission of data from transmitter to receiver

A. Transmitter

At the transmitter, initially we generate data randomly by the source and the generated data is fed to LDPC encoder. LDPC Encoder does two tasks, which are, creation of LDPC matrix and generation of parity bits. 

Creation of LDPC matrix:

There are two methods for distributing a non-zero element. One method is by placing 1’s uniformly at random along each column and the other is by placing 1’s uniformly at random along each column and row. Try to eliminate girth-4 because of looping problems. Create full sparse matrix \( T \) by employing one of the above methods. The created matrix will be of size \( M \times N \) where, \( M \) represents the number of parity check bits and \( N \) represents the block length. Hence, \( R=1/2 \) LDPC matrix is created.

Generation of Parity bits:

Using sparse matrix properties of \( T \) and sparse LU decomposition parity bits are computed. Let \( T = [D|E] \). Now, we have to decompose \( D \) to \( LU \) where, \( L \) represents Lower triangular matrix and \( U \) represents Upper triangular matrix. To decompose the matrix \( D \), it must be non-singular. Hence 1’s must be reordered diagonally. Thereby we get parity-check bits and the reordered sparse matrix \( G \), which is used for encoding and decoding instead of original sparse matrix \( T \).

We can use any of the modulation schemes such as BPSK, QPSK, QAM, and so on. Here, we have used BPSK modulation and applied modulation scheme to the LDPC encoded bits.

To the modulated bits, apply Block Diagonalization technique which is a combination of SVD technique and Water Filling algorithm [7].

SVD technique:

SVD is performed on the channel matrix such that channel matrix \( H \) can be decomposed into \( S, V, D \) matrices [8].

\[ [A, B, C] = SVD(H) \]  

(2.1)

where, \( H \) is channel matrix and is equal to 
\[ \begin{bmatrix} \tilde{d}_{11} & \tilde{d}_{12} \\ \tilde{d}_{21} & \tilde{d}_{22} \end{bmatrix} \], 
\( A \) and \( C \) are unitary matrices, 
\( B \) is a diagonal matrix.

Apply Water Filling algorithm.

Water Filling Algorithm: [9], [10]

1. Consider absolute value of noise vector at different frequencies in linear units. Noise vector must be a row vector.
2. Also consider power constraint and acceptable tolerance in the same units of noise vector.
3. Initial water level should be considered in the units of noise vector such that abs (Power Constraint - SUM(MAX(Initial Water Level – Noise Vector, 0))) <= Acceptable Tolerance
4. Build it in such a way that Power Constraint >= SUM (MAX (Initial Water level, 0)) and it should be never less than SUM (MAX (Initial Water level, 0))

Modulated bits are multiplied with Vector \( C \), Power Constraint and then transmitted through the channel using Space time block codes.

STBCs means transmitting copies of data i.e., in the first time slot the information will be transmitted as \([p_1, p_2]\) and in the second time slot transmit the information as \([-p_1^*, p_2^*]\). Here, \( p_1 \) and \( p_2 \) are the data bits which are obtained after the multiplication of \( C \), power constraint with modulated bits [10], [11]. As we are transmitting copies of data, we can recover the original signal, thereby reducing signal fading effects.

Algorithm at transmitter:

- By a random source generate a stream of data.
- Create an LDPC matrix.
- Generate parity check bits. These two steps constitute LDPC encoder and we get LDPC encoded bits.
• Apply the modulation technique to the encoded bits.
• By considering channel state information is known, apply the Singular Value Decomposition technique to the channel matrix.
• Apply Water Filling algorithm. These two steps constitute Block Diagonalization technique.
• Multiply the data with a unitary matrix obtained from SVD technique and Power Constraint obtained from Water Filling algorithm.
• Transmit the modified data through the channel with the help of STBCs.

B. Receiver
At the receiver antenna, bits received in two time slots are as follows:

\[ Y_1 = [H][P_1] + [N_1] \] (2.2)
\[ Y_2 = [H][P_2] + [N_2] \] (2.3)

where, \( H \) is the channel matrix, \( P_1 \) and \( P_2 \) represents the input bits to the channel of size \( N_t \times b \) in first and second time slots respectively, \( N_t \) represents number of transmitter antennas and \( b \) represents number of bits after modulation.

[Here, \( N_t \) and \( N_1 \) represents receiver noise matrices in time slots 1 and 2 respectively and are of size \( N_t \times b \).]

As the first step at receiver, apply MMSE detection to the received bits. It is performed by [15]

\[ Z_1 = ((H_1^H H_1) + (\sigma^2 I))^{-1} H_1^H Y_1 \] (2.4)
\[ Z_2 = ((H_1^H H_1) + (\sigma^2 I))^{-1} H_1^H Y_2 \] (2.5)

Here, \( H_1 = HC \), \( H \) is the Channel Matrix, \( C \) is right hand unitary matrix obtained from SVD technique, \( \sigma^2 \) is the variance, \( I \) is the Identity Matrix, \( Y_1 \) and \( Y_2 \) are the matrices obtained from equations (2.2) and (2.3) respectively, \( Z_1 \) and \( Z_2 \) are the bits estimated in first time slot and second time slot respectively.

For the obtained output matrices \( Z_1 \) and \( Z_2 \) perform the corresponding demodulation according to the modulation scheme applied.

For the demodulated bits apply LDPC decoding based on Sum-Product Algorithm (SPA). There are 4 versions of SPA decoder which are

- Hard-Decision decoder – It is also known as Bit-Flip decoder. It decodes 1/0 message as follows. If the majority of bits present is 1 then it chooses ‘1’ or else ‘0’. The decoder could be used for introduction to message passing algorithms since it does not employ complicated probability. Expect worse performance compared for very low Eb/N0.

- Probability-Domain SPA decoder – Based on Gallager’s work

- Log-Domain SPA decoder – Similar to probability-domain SPA, but instead of probability function, use log-likelihood. Advantage is that operations can be performed using additions instead of multiplications, which takes less computational time.

- Simplified Log-Domain SPA decoder – It is similar to Log-Domain SPA decoder, but log-likelihood function is directly replaced by incoming signal waveform. Hence simplified log-domain SPA decoder doesn’t need noise variance information.

Here, we used Log-Domain SPA decoder and performed decoding process. After the decoding process, calculate BER. Performance of the proposed technique is obtained by repeating the above steps with various SNR.

III. RESULTS
By assuming that we have channel state information and by creating random LDPC codes, results are obtained as follows:

Fig. 3.1: Performance comparison of MMSE and ZF detectors for 2×2 MIMO systems

From fig. 3.1 we observe that at 6dB SNR, BER for ZF is \( 2 \times 10^{-3} \) and BER for MMSE is \( 6 \times 10^{-3} \). From the graph we conclude that the...
Performance of MMSE is better than ZF as it has less BER when compared with ZF. Hence, we use MMSE detection technique instead of ZF detection technique.

**Fig -3.2:** Performance comparison of TCM and LDPC codes for 2×2 MIMO systems

From fig-3.2 we observe that at 1dB SNR, BER for TCM encoded MMSE is nearly $10^{-2}$ and for LDPC encoded MMSE is nearly $10^{-3}$. Hence, we conclude that LDPC performance is superior to TCM. So, LDPC codes are encoded to Block Diagonalization technique.

**Fig -3.3:** Performance comparison of MMSE and BD for 2×2 MIMO systems

From fig-3.3 we observe that at 2dB SNR, BER for MMSE is $2 \times 10^{-1}$ and for BD is $4 \times 10^{-2}$. So, from the graph we conclude that performance of BD is better than MMSE technique. Hence, we applied LDPC encoding to Block Diagonalization technique.

**Fig -3.4:** Performance comparison of BD with and without LDPC codes for 2×2 MIMO systems

From fig-3.4, we observe that at 2dB SNR, BER for Block Diagonalization technique without any encoding schemes is $5 \times 10^{-2}$ and for LDPC encoded Block Diagonalization technique is nearly $2 \times 10^{-3}$. Hence, LDPC encoded Block Diagonalization technique is superior to BD technique in performance.

**Fig -3.5:** Performance comparison of BD-STBC with and without LDPC encoding for 2×1 MIMO systems

From fig-3.5, we observe that performance of LDPC-BD-STBC is superior to BD-STBC because of the application of LDPC encoding. At 1dB BER for LDPC-BD-STBC is $10^{-2}$ and for BD-STBC is $7 \times 10^{-3}$. BD-STBC technique works well in Rank Deficient scenario and by the application of LDPC encoding its performance is improved even better. Hence, our proposed technique works well in practical Rank Deficient scenario also where, most of the receivers fail.

**IV. CONCLUSION**

LDPC is an efficient encoding technique when compared with TCM, BD reduces the effect of MU and STBCs reduces signal fading effects. In this paper, we proposed LDPC aided Block Diagonalization technique using STBCs (LDPC-BD-STBCs) which works better in rank deficient scenarios when compared to TCM-BD-STBCs. Bit error rate must be reduced in the Rank Deficient Scenario. This scenario is present when number of transmitters is greater than number of receivers. Most of the techniques such as ZF and MMSE fail in this scenario. But from the results it is proved that LDPC-BD-STBC technique achieves better performance even in this scenario.
REFERENCES


15. Meiyuan Ju, Jia Qian, Yueheng Li, Guoping Tan and Xujie Li, IEEE Student Member, “Comparison of Multiuser MIMO Systems with MF, ZF and MMSE Receivers”, Third International Conference on Information Science and Technology March 23-25, 2013; Yangzhou, Jiangsu, China