SIGNAL DETECTION FOR SPATIALLY MULTIPLEXED MIMO SYSTEM

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Abstract

In this paper; we are studying techniques for signal detection in Multiple Input Multiple Output (MIMO) spatial multiplexing system. This paper includes techniques like Maximum likelihood (ML). The ML detection achieves the optimal performances with reduced complexity using BPSK modulation. The simulation result of ML decoder shows the Bit error rate (BER) reduces if signal to noise to ratio increases.

Keywords: MIMO, Spatial Multiplexing, Detection technique, Maximum likelihood detection, Equalization based

detection technique.

1. INTRODUCTION

MIMO system can defined as point to point communication having multiple numbers of antennas at the transmitter and receiver side. By using multiple antennas at both sides clearly gives the performance enhancement over diversity systems. In recent research it is observed that MIMO can drastically increase the data rates of systems with no increase in the transmitting power and bandwidth of the system [1] [2]. Multiple inputs multiple output channels gives many advantages compared to Single Input Single Output (SISO) channels like array gain, the diversity, and the multiplexing gain. MIMO can yield significantly improve reliability of link and increase data rate without using additional bandwidth. The unique characteristic of MIMO channels is multiplexing gain. Spatial multiplexing is a multiplexing technique in MIMO to transmit separate and independent data signals is also called as streams, from each of the different transmitting antenna. Thus, spatial multiplexing is powerful technique which increases channel capacity at higher SNR values [3-5]. There are numerous detection techniques in MIMO wireless communication such as Maximum Likelihood detection (ML); Equalization based detection, sphere projection algorithm.

In this paper, we propose a ML detection technique using BPSK modulation. Results of simulation demonstrate significant improvement in performance and BER. The main advantage of the proposed technique is increasing the performance of received signal in order to maintaining the fixed and less computational complexities.

2. SYSTEM MODEL



Fig 1: Block Diagram of MIMO Detection technique

Figure1 shows the block diagram MIMO detection technique. Detection is technique of extracting original signal from received signal. Maximum likelihood; Equalization based detection is used in proposed system. In MIMO high data rate is achieved by the mean of spatial multiplexing technique, in which Independent stream of information is transmitted in parallel over the different transmitting antenna [6]. Here, we consider fading channel in MIMO with N_T antenna for transmitting, N_R antennas at receiving and N_R \geq N_T. In MIMO spatial multiplexing system n^{th} information symbol b_{m} is transmitted directly on n^{th} transmit antenna. At a time instant this gives a baseband model with

$$\mathbf{v} = \mathbf{H} \mathbf{b} + \mathbf{c} \tag{1}$$

The N_T×1 transmit vector $\mathbf{b} = (\mathbf{b}_1 \ \mathbf{b}_2 \ \dots \ \mathbf{B}_{\text{NT}})^{\text{T}}$, The channel matrix H is N_R×N_T the N_R×1 received vector $\mathbf{v} = (\mathbf{v}_1 \ \mathbf{v}_2 \ \dots \ \mathbf{v}_{\text{NT}})^{\text{T}}$, and N_R×1 noise vector $\mathbf{c} = (c_1 \ c_2 \ \dots \ \mathbf{C}_{\text{NR}})^{\text{T}}$. The (n, m)th entry of H, $H_{n,m} = (H)_{n,m}$, is complex fading coefficient between the the transmitting and receiving antenna. The b_m is data symbol taken from a complex valued symbol A and is considered to be mean zero and independent with unit variance. The c_m is noise and it is considered to be symmetric in circularly with σ_w^2 variance. At the receiver side, channel H is considered to be purely known [6].

The diversity in a MIMO SM system can be shown by the number of receive antenna $N_{\rm R}$. if $H_{\rm n,m}$ channel coefficient is independent Maximum diversity is available, because each information data b_m is transmitted over $N_{\rm R}$ independent scalar fading channels $H_{n,m}$, $n = 1, ..., N_{\rm R}$ [6]. If larger is $N_{\rm R}$, smaller is the possibility that all of channels fade at a time and thus the data detection reliability can be increased. If N_R available diversity, than the symbol error rate of Maximum likelihood detector technique decays like SNR^{-M_R} in the high regime [7].

2.1 Maximum Likelihood Detection

Maximum likelihood is a perfect detection technique in terms of less error and it fully uses the available diversity. From system model (1) the maximum likelihood detection can be given by

$$d_{ML} = \arg \min_{b \in D} \{ \| v - Hb \|^2 \}.$$
 (2)

Here, $\mathcal{D} = \mathcal{A}^{NT}$ indicates the set of possible transmitted information vectors b. The cardinality of $|\mathcal{D}| = |\mathcal{A}^{NT}|$ and hence grows exponentially with N_T [8] [9]. The maximum likelihood receiver tries to minimize \hat{x} which minimizes,

$$L = |d - H\hat{s}|^{2}$$
(3)
$$\left[\begin{pmatrix} d_{1} \\ d_{2} \end{pmatrix} - \begin{pmatrix} h_{1,1} & h_{1,2} \\ h_{2,1} & h_{2,2} \end{pmatrix} \begin{pmatrix} \hat{s}_{1} \\ \hat{s}_{2} \end{pmatrix} \right]^{2}$$
(4)

The simulated result include maximum likelihood detection in which main aim is to find the minimum among all the transmitted symbol based on the minimum chose that is estimate of transmitted data and it is repeated for more number of values of E_b/N_0 [10].

2.2 Equalization Based Detection

In equalization detection, an estimation of transmitted information b is made as d = Gv with an "equalization matrix" G. The detected information vector is obtained as $\hat{b} = Q\{d\}$, where Q $\{\cdot\}$ indicate the component wise according to the symbol \mathcal{A} [6]. For the *zero-forcing* equalizer, the equalization matrix G is given by pseudo-

inverse of H, it is given by $G = H^{\#} = (H^{H}H)^{-1}H^{H}$. Thus, zero forcing equalization result is given by

$$y_{ZF} = H^{\#} r = (H^{H} H)^{-1} H^{H} r = d + \widetilde{w}$$
,

The transmitted information vector b is lost by the the noise in channel $\tilde{c} = H^H c$. This results the information caused by channel H is completely removed. The zero-forcing equalization received vector can be used as solution to the maximum likelihood detection problem, where the information set \mathcal{D} underlying ML detection is relaxed to the convex set $\mathbb{C}^{M_T}[9]$.

$$y_{ZF} = arg \min_{y \in \mathbb{C}^{N_T}} \{ \| v - Hd \|^2 \}.$$
 (5)

The noise effect in the ZF equalizer can be decreased by using the MMSE equalizer $(H^H H + \sigma_w^2 I)^{-1} H^H r$, and G minimizes the mean square error E { $\| Gv - b \|^2$ }, the MMSE equalization is given by $y_{MMSE} = (H^H H + \sigma_w^2 I)^{-1} H^H r$ (6)

3. RESULT ANOD DISCUSSION

The Maximum likelihood detection tries to find the \hat{x} in equation (2). The simulation ML decoder mainly include the finding the minimum among the four possible transmitted Symbols, based on the minimum chose the estimate of the transmitted symbol. This is repeated for no of values of Eb/N0 and plot the simulated result and theoretical results.

The figure 2 show performance of MIMO Maximum likelihood detection in terms of BER and SNR for 1x2 and 2x2 transmit diversity system. As per the graph BER decreases with increase in the SNR. At BER $\sim 10^{-3}$, the SNR values ~ 11 dB for both the cases 1x2 and 2x2 transmit diversity and shows that the MIMO-ML detection shows same performance as those of 1x2 transmit diversity system.



4. CONCLUSION

The ML Detection is used to extract original signal from the received signal in MIMO. The Proposed ML detection gives the performance improvement compared to the conventional detection like equalization based and sphere decoding. The simulation result of $2x^2$ transmit diversity is close to the $1x^2$, It shows the bit error rate (BER) decreases as with the increase in the signal to noise ratio.

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