AN EFFICIENT COMPARISON OF MIMO OFDM CHANNEL ESTIMATION ALGORITHM

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Abstract

A multiple-input multiple-output (MIMO) wireless communication system combined with the orthogonal frequency division multiplexing (OFDM) modulation technique, which can achieve reliable high data rate transmission over broadband wireless channels. Channel state information for MIMO Channel systems based on pilot aided arrangement is included in this paper. The estimation of channel at pilot frequencies with conventional Least Square (LS) and Minimum Mean Square (MMSE) estimation algorithms is carried out through Matlab simulation. The performance of MIMO OFDM is evaluated on the basis of Bit Error Rate (BER) and Mean Square Error (MSE) level. Further enhancement of performance can be achieved through maximum diversity Space Time Block Coding (STBC) and Maximum Likelihood Detection at transmission and reception ends respectively. MMSE estimation has been shown to perform much better than LS for the MIMO system using pilot carriers.

Keywords

LS, MMSE, DFT, MIMO OFDM, Channel Estimation, MIMO-OFDM, Pilot carriers, Diversity, Spatial Multiplexing, Space time coding
Introduction

OFDM (Orthogonal Frequency Division Multiplexing) is becoming a very popular multi-carrier modulation technique for transmission of signals over wireless channels. OFDM divides the high-rate series stream into parallel lower rate data and which helping to eliminate Inter Symbol Interference (ISI). It also allows the bandwidth of Subcarriers to overlap without Inter Carrier Interference (ICI) as long as the modulated carriers are orthogonal. OFDM therefore is considered as an efficient modulation technique for broadband access in a dispersive environment.[1] In this new information age, Wireless systems continue to make every effort for ever higher data rates. This goal is particularly challenging for systems that are power, bandwidth, and complexity limited. High data rate and strong reliability systems are becoming the dominant factors for a successful exploitation of commercial networks in wireless communication. MIMO-OFDM (multiple input multiple output orthogonal frequency division multiplexing), a new wireless broadband technology, has gained great popularity for its capability of high rate transmission, reliability and its robustness against multi-path fading and other channel impairments. The arrangement of multiple
antennas at the transition end and reception end results increase in the diversity gain refers the quality of signal and multiplexing gain refers the transmission capacity. Space time block coding used in this paper to transmit multiple copies of a data stream across a number of antennas and to exploit the various received versions of data to improve reliability of data transfer. The major challenge faced in MIMO-OFDM systems is how to obtain the channel state information accurately and promptly for coherent detection of information symbols. In the graph we have shown the advantage of channel estimation compared to that of no channel estimation for the different SNR levels.

The training-based method channel estimation can be performed by either block type pilots where pilot tones are inserted into all frequency bins within periodic intervals of OFDM blocks or by comb pilots where pilot tones are inserted into each OFDM symbol symbols with a specific period of frequency bins. In an OFDM system, the transmitter modulates the message bit sequence in to PSK/QAM symbols, performs IFFT on the symbols to convert them in to time domain signals, and sends them throughout the wireless channel. The channel characteristic distort the received signal, so to recover the transmitted bits at receiver side, the channel effect must be estimated and compensated. (8-10). Each subcarrier can be regulated as an independent channel, as long as no inercarrier interference occurs, and preserved orthogonality among subcarriers. The orthogonality allows each subcarrier component of the received signal to be expressed at the subcarriers. So the transmitted signal can be recovered by estimating the channel response just at each

\[ \text{MIMO channel model} \]
Fig. 1 depicts a block diagram of the MIMO OFDM system. We consider MIMO–OFDM systems with two transmit antennas and two receive antennas. The total number of subcarriers is $N$. Basically, the MIMO-OFDM transmitter has $N_t$ parallel transmission paths which are very similar to the single antenna OFDM system, each branch performing serial-to-parallel conversion, pilot insertion, $N$-point IFFT and cyclic extension before the final TX signals are up-converted to RF and transmitted. It is worth noting that the channel encoder and the digital modulation, in some spatial multiplexing systems, can also be done per branch, where the modulated signals are then space-time coded. Subsequently at the receiver side, the CP is removed and $N$-point FFT is performed per receiver branch. Next, the transmitted symbol per TX antenna is combined and outputted for the subsequent operations like digital demodulation and decoding. Finally all the input binary data are recovered with certain BER. As a MIMO signalling technique, $N_t$ different signals are transmitted simultaneously over $N_t \times N_r$ transmission paths and each of those $N_r$ received signals is a combination of all the $N_t$ transmitted signals and the distorting noise. It brings in the diversity gain for enhanced system capacity as we desire[7].

MIMO OFDM system model which is consist of $x_t$ \(t=0,...,N-1\), transmitted signals and $y_t$ received signals. The transmitted signals $x_t$ are taken from multi amplitude signal constellation The channel impulse response of the system is calculated using the equation given below.

$$g(t) = \sum (t-\tau) \quad \ldots \quad (1)$$

Where $\tau$ is the sampling interval, $d$ is the delay, $a$ is the amplitude.

The received signal is given as

$$Y = XFg + n$$

Where $X$ is the matrix element of $x$ on its diagonal and $n$ is the noise.

$$n = \sqrt{F^2} \quad \ldots \quad (2)$$

Training Symbol-Based Channel Estimation

Training symbols can be used for channel estimation, usually providing a good performance. However, their transmission efficiencies are reduced due to the required overhead of training symbols such as preamble or pilot tones that are transmitted in addition to data symbols. The least-square (LS) and minimum-mean-square-error (MMSE) techniques are widely used for channel estimation when training symbols are available [9]. We assume that all subcarriers are orthogonal. Then the training symbols for $N$ subcarriers can be represented by Matrix,

$$X = [\ldots]$$

Where $X[k]$ denotes a pilot tone at the $k$th subcarrier with $E\{X[k]\} = 0$ and $Var\{Z[k]\} = \sigma_z^2, k=0,1,2\ldots,N-1$. Channel gain
is $H[k]$ for each subcarrier $k$, the received training signal $y[k]$ can be represented by,

$$Y = XH + Z$$

Where $H$ is a channel vector given as $H = [H[0], H[1], \ldots, H[N-1]]^T$ and $Z$ is a noise vector given as $Z = [Z[0], Z[1], \ldots, Z[N-1]]^T$ with $Var\{Z[k]\} = \sigma^2_z$, $k=0,1,2,\ldots,N-1$.

**LS Channel Estimation**

The least square channel estimation method finds estimate $\hat{H}$ in such a way that the following function can be minimized.

$$J(\hat{H}) = ||\hat{H}||^2 = \hat{H}^H\hat{H} = Y^HY - Y^HX\hat{H} + \hat{H}^HX^HY + \hat{H}^HX^HX\hat{H}$$

derivation of the function with respect to $\hat{H}$ to zero,

$$\frac{\partial J}{\partial \hat{H}} = -2(XHY) + 2(X^HX\hat{H})^* = 0 \quad \text{………(5)}$$

we have $X^HX\hat{H} = X^HY$, which gives the solution to the LS channel estimation as,

$$\hat{H}_{LS} = (X^HX)^{-1}X^HY = X^{-1}Y \quad \text{………(6)}$$

$\hat{H}_{LS} = \quad ,K=0,1,2,3,\ldots,N-1$

The mean square error (MSE) of this LS channel estimation is given as

$$\text{MSE}_{LS} = E\{ (H-\hat{H}_{LS})^H(H-H_{LS}) \} = E\{ (H-X^{-1}Y)^H(H-X^{-1}Y) \} = E\{ X^{-1}Z^H(X^{-1}Z) \} = E\{ Z^H(XX^{-1}X)^{-1}Z \} \quad \text{………(7)}$$

The MSE of LS in above equation is inversely proportional to the $\text{SNR} = \frac{\text{SNR}}{\text{SNR} + \sigma^2_z}$, which implies that it may be subject to noise enhancement. Due to its simplicity the LS is widely used for channel estimation.

**MMSE Channel estimation**

In MMSE channel estimation weight matrix $W$ define $\hat{H}$, refer fig 2 MSE of the channel estimate $\hat{H}$ is given as,

$$J(\hat{H}) = E\{ ||\hat{H}||^2 \} = \hat{H}^H\hat{H} = Y^HY - Y^HX\hat{H} + \hat{H}^HX^HY + \hat{H}^HX^HX\hat{H}$$

Then the MMSE Channel Estimation method finds a better linear estimate in terms of $W$ in such a way that the MSE is minimized.

The orthogonality principal state that the estimation error vector $e = H - \hat{H}$ is orthogonal to $\sim$,

$$E\{ e^H \} = E\{ (X^HX)^{-1}X^HY - \hat{H}^H \} = E\{ H^H \} - WE\{ \sim^H \} \quad \text{………(9)}$$

$\text{R}_{AB}$ is the auto correlation matrix of $N \times N$ matrices of A and B

$W = \quad \text{………(10)}$

Where $R^-$ is the auto correlation matrix of
is given as
And $R_{H}$ is the crosscorrelation matrix between the true channel vector and temporary channel estimate vector in the frequency domain.

$\tilde{H} = W = R_{H} - R^{-1}$

Simulation results and graphs

1) No channel estimation vs. LS channel estimation

<table>
<thead>
<tr>
<th>FFT SIZE</th>
<th>SNR</th>
<th>BER</th>
</tr>
</thead>
<tbody>
<tr>
<td>128</td>
<td>16</td>
<td>0.02375</td>
</tr>
<tr>
<td>256</td>
<td>16</td>
<td>0.1542</td>
</tr>
<tr>
<td>512</td>
<td>16</td>
<td>0.015</td>
</tr>
<tr>
<td>1024</td>
<td>8</td>
<td>0.000833</td>
</tr>
</tbody>
</table>

$\tilde{R} = E\{\tilde{H}\} = E\{X^{-1}Y(X^{-1}Y)^H\}
= E\{(H+X^{-1}Z)(H+X^{-1}Z)^H\}
= E\{HH^H+X^{-1}ZHH^H+HZ^H(X^{-1})^H+X^{-1}ZZ^H(X^{-1})^H\}
= E\{HH^H\}+E\{X^{-1}ZZ^H+HZ^H(X^{-1})^H\}
= E\{HH^H\}+I \quad \ldots \quad (11)$

$\tilde{R} = R_{H}(R_{HH} + \sigma_x^2 / \sigma_z^2 I)^{-1} \quad \ldots \quad (12)$

LS vs MMSE based on MSE(mean Square Error) different values of FFT
<table>
<thead>
<tr>
<th>FFT SIZE</th>
<th>SNR</th>
<th>MSE(LS)</th>
<th>MSE(MMSE)</th>
</tr>
</thead>
<tbody>
<tr>
<td>64</td>
<td>15</td>
<td>0.1563</td>
<td>0.03133</td>
</tr>
<tr>
<td>128</td>
<td>15</td>
<td>0.1704</td>
<td>0.02208</td>
</tr>
<tr>
<td>256</td>
<td>15</td>
<td>0.04077</td>
<td>0.007673</td>
</tr>
</tbody>
</table>

for FFT=1024 too much long time.

LS vs MMSE based on SER (symbol error rate) different value of FFT
See command window for detail values.
Reading for 512 FFT size and computation time almost 1:30 Hr

ser_ls = 0.4955
ser_mmse = 0.4947
ser_ls =
0.4955 0.4922
ser_mmse = 0.4947 0.4913
ser_ls = 0.4955 0.4922 0.4895
ser_mmse = 0.4947 0.4913 0.4884
ser_ls = 0.4955 0.4922 0.4895 0.4874
ser_mmse = 0.4947 0.4913 0.4884 0.4872
ser_ls = 0.4955 0.4922 0.4895 0.4874 0.4870
ser_mmse = 0.4947 0.4913 0.4884 0.4872 0.4861
ser_ls = 0.4955 0.4922 0.4895 0.4874 0.4870
ser_mmse = 0.4947 0.4913 0.4884 0.4872 0.4861 0.4870
Conclusion

In this paper, we propose to evaluate the performance of LS and LMMSE estimation techniques. For Downlink systems under the effect of the IFFT size, channel length, SNR and BER. The cyclic prefix inserted at the beginning of each OFDM symbol is usually equal to or longer than the channel length in order to suppress ICI and ISI. However, the CP length can be shorter than the channel length because of some unforeseen behavior of the channel. Simulation results show that LS channel estimation is better than that of NO channel estimation. When LS is compared to MMSE estimation MMSE shows better performance than LS. As the IFFT size increased the significant improvement in BER but increasing the IFFT size will result in more computational load. LS is easy to implement and MMSE is a bit complex but gives significant increase in SNR value when compared to LS length, the LMMSE performs better.

REFERENCES


