

# Minimum Probability of Error Demodulation for MC-CDMA

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**Abstract**—Multi-Carrier Code Division Multiple Access (MC-CDMA) is an attractive choice for high speed wireless communication as it mitigates the problem of Intersymbol Interference (ISI) and also exploits frequency diversity. The data stream in MC-CDMA is spread using a user specific spreading sequence in the frequency domain and every chip travels on a different sub-carrier. At the receiver, the different sub-carriers have to be combined properly to form the decision variable. Minimum Mean Square Error (MMSE) detector is a popular choice for this, but as it tries to minimize the Mean Square Error (MSE) and not the Bit Error Rate (BER), the BER it attains is not optimum. In this paper, we propose a Minimum Probability of Error (MPOE) based linear multiuser detector, which tries to minimize the probability of error, by exploiting the probability density function (pdf) of the filter output. Theoretical and simulation results are presented for the MPOE detector and it can be seen that the MPOE detector outperforms the MMSE detector.

**Index Terms**—MPOE, MMSE, MC-CDMA, Multiuser Detection

## I. INTRODUCTION

Multi-Carrier Code Division Multiple Access (MC-CDMA) is a fairly recent multiple access technique that has attracted much research during the past decade. A combination of Orthogonal Frequency Division Multiplexing (OFDM) and CDMA [1], MC-CDMA combines the advantages of both the schemes. Like OFDM, MC-CDMA signals have attractive spectral characteristics. Their spectral energy is confined almost to the allocated bandwidth. Owing to this and other spectral characteristics, MC-CDMA allows for easier system coexistence. Moreover, by transmitting the same data bit over different sub-carriers, MC-CDMA exploits frequency diversity better than OFDM and also reduces the symbol rate on each of the sub-carrier. Due to this reduced symbol rate, the effect of Intersymbol Interference (ISI) on the transmitted signals is also reduced. The residual ISI can be eliminated using sufficient guard interval. The data bits of the different users are spread using a user specific spreading waveform in the frequency domain.

This paper focuses on the uplink scenario, where the mobile sets communicate with the base station. The base station receives a composite signal from all the users active in that cell. Multiuser detection deals with the problem of the reliable demodulation of the user's bits in such a scenario. Various detectors, such as MRC, ERC and MMSE [2] have already

been proposed for the uplink case. Traditionally, adaptive filtering in communications has been developed based on the Mean Square Error (MSE) criterion [4]. Recently, [5] - [8], several linear detectors have been suggested for CDMA and OFDM based on the principle of Minimum Probability of Error (MPOE). As these detectors try to minimize the BER and not the MSE, these detectors can achieve a better BER than the MMSE detectors.

In this paper we propose a detector for uplink MC-CDMA based on the MPOE criterion. The theoretical and simulation results for the same have been presented. It can be seen that the MPOE detector outperforms the MMSE detector in terms of BER albeit at a slightly higher computational complexity when in weight adaptation mode. Once in the decision directed mode, both MMSE and MPOE are linear detectors. Ways to reduce the computational complexity of the MPOE detectors in the weight adaptation mode have also been proposed in the paper. Also another advantage of the MPOE detector over MMSE detector is its improved resistance to the Near Far Problem prevalent in mobile communication.

The organization of the paper is as follows. Section II describes the signal model assumed. Section III derives the probability of error for MC-CDMA and arrives at the MPOE detector. Simulation results are presented in Section IV. The conclusions are presented in Section V.

## II. SIGNAL MODEL

In a MC-CDMA system, the incoming data bits are multiplied by the user specific spreading sequence and then each chip is transmitted on a different sub-carrier. The sub-carrier spacing is  $\frac{F}{T_b}$ , where  $F$  is an integer greater than or equal to one. If  $F = 1$ , then the system conserves bandwidth. If  $F > 1$ , then the system uses more bandwidth, but at the same time achieves better frequency diversity. The spectra of the various sub-carriers overlap, but the information carried by the different sub-carriers can still be recovered due to the orthogonality of the sub-carriers. An efficient way of implementing the modulation digitally is to perform an Inverse Discrete Fourier Transform (IDFT) on the incoming symbols and then D/A convert the sequence obtained. The resulting analog waveform can be upconverted and then transmitted.

If the bit vector corresponding to the  $N$  users is written as

$$\mathbf{A} = [a_0 \ a_1 \ \dots \ a_{N-1}]^T \quad (1)$$

where  $a_i \in \{+1, -1\}$  if user  $i$  is active and 0 otherwise and the orthonormal code matrix,

$$C = \begin{bmatrix} c_0[0] & c_1[0] & \dots & c_{N-1}[0] \\ c_0[1] & c_1[1] & \dots & c_{N-1}[1] \\ \dots & \dots & \dots & \dots \\ c_0[N-1] & c_1[N-1] & \dots & c_{N-1}[N-1] \end{bmatrix} \quad (2)$$

where the  $i^{\text{th}}$  column corresponds to the spreading sequence of the  $i^{\text{th}}$  user, then the discrete time transmitted sequence can be written as

$$S = [s_0 \ s_1 \ \dots \ s_{N-1}]^T = F^{-1}CA \quad (3)$$

where  $F^{-1}$  is the  $N$  point IDFT matrix. The actual analog waveform during the  $k^{\text{th}}$  symbol duration that is transmitted after upconversion is

$$s_k(t) = \sum_{m=0}^{N-1} \sum_{i=0}^{N-1} c_m[i] a_m[k] \cos(2\pi f_c t + 2\pi i \frac{F}{T_b} t) \quad (4)$$

The effect of the channel on the transmitted signal can be written as

$$y[n] = (h * s)[n] + \eta, \quad 0 \leq n \leq N-1 \quad (5)$$

where  $h[l]$ ,  $0 \leq l \leq L-1$  is the discrete time channel impulse response and  $\eta$  is the additive white Gaussian noise (AWGN). Because of the linear convolution of the transmitted signal with the channel impulse response, the received signal is affected by ISI and Interchannel Interference (ICI). Equation (5) can be written as [3]

$$\begin{aligned} y_i[n] &= \sum_{l=0}^{L-1} h[l] s_i[n-l]_N \sigma[n-l] \\ &+ \sum_{l=1}^{L-1} h[l] s_{i-1}[n-l]_N (1 - \sigma[n-l]) + \eta \\ &= y_i^{(i)}[n] + y_i^{(i-1)}[n] + \eta \end{aligned} \quad (6)$$

where  $y_i[n]$  is the received signal in the  $i^{\text{th}}$  symbol duration and  $y_i^{(i)}[n]$  and  $y_i^{(i-1)}[n]$  are the contributions to it from the symbols  $s_i[n]$  and  $s_{i-1}[n]$ , transmitted in the  $i^{\text{th}}$  and  $(i-1)^{\text{th}}$  duration respectively.  $\sigma$  is the Heaviside function and  $[n]_N$  is  $n$  modulo  $N$ . It can be seen that due to the multipath, the first  $L-1$  symbols of received signal are distorted by ISI from the previous symbol. The received signals after the DFT can be written as

$$Y_i = H_{ICI} S_i + H_{ISI} S_{i-1} + N \quad (7)$$

where

$$H_{ICI}(k, m) = \sum_{l=0}^{L-1} H_l^i(m-k) e^{-j \frac{2\pi}{N} l m} \quad (8)$$

$$H_l^i(m-k) = \frac{1}{N} \sum_{n=0}^{N-1} h[l] \sigma[n-l] e^{j \frac{2\pi}{N} n(m-k)}$$

and

$$H_{ISI}(k, m) = \sum_{l=1}^{L-1} H_l^{i-1}(m-k) e^{-j \frac{2\pi}{N} l m} \quad (9)$$

$$H_l^{i-1}(m-k) = \frac{1}{N} \sum_{n=0}^{N-1} h[l] (1 - \sigma[n-l]) e^{j \frac{2\pi}{N} n(m-k)}$$

$H_{ICI}$  is the ICI matrix and  $H_{ISI}$  is the ISI matrix.  $S_i$  and  $S_{i-1}$  are the  $N$  point DFT of  $s_i[n]$  and  $s_{i-1}[n]$  respectively.  $N$  is the Gaussian noise vector. In the downlink case that we are considering, we cannot eliminate this ISI because we do not perform a joint demodulation due to computational constraints in a mobile set. Therefore in order to avoid ICI and ISI, sufficient cyclic prefix (CP) has to be added. After the introduction of cyclic prefix of sufficient length, the linear convolution of equation (5) starts looking like a circular convolution. Also a circular convolution in the time domain is equivalent to a point wise multiplication of the DFT's in the frequency domain. In such a case, the ISI matrix turns out to be a zero matrix and ICI matrix is a diagonal matrix with  $H_{ICI}(k, k) = \text{DFT}\{h[l]_k\}$ . Thus the effect of the channel is now an independent frequency flat fading for each of the sub-carrier. The received signal in this case can be written as

$$Y = HCA + N \quad (10)$$

where

$$H = \begin{bmatrix} \rho_0 e^{j\theta_0} & \dots & 0 & 0 \\ 0 & \rho_1 e^{j\theta_1} & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & \rho_{N-1} e^{j\theta_{N-1}} \end{bmatrix} \quad (11)$$

$\rho'_i s$  are I.I.D Rayleigh random variables and  $\theta'_i s$  are uniformly distributed random variables.

### III. MINIMUM PROBABILITY OF ERROR DETECTOR

At the transmitter the different users are separated using orthogonal Walsh Hadamard codes but due to the independent scaling of each of the sub-carriers, the code orthogonality between the users is destroyed. After the DFT is performed on the received signal, a single tap equalizer has to restore this orthogonality before forming the decision variable. The MMSE receiver tries to minimize the mean square error between the transmitted and the estimated symbol. Since the BER, and not the MSE is the deciding factor in a digital communication system, a detector which tries to minimize the BER is desired. The probability density function (pdf) of

the output of the linear filter acting on  $Y$  is a mixed sum of Gaussian and it is seen that a MPOE based detector can better exploit this pdf and achieve a lower BER. In the following sections we will derive the expression for the probability of error for MC-CDMA and then use this expression to arrive at the MPOE detector.

#### A. Probability of Error

Let  $w = [w_0 \ w_1 \ \dots \ w_{N-1}]^H$  be the filter vector used to demodulate the bits of the desired user (user 0 in our case). The filter output  $x$  can be written as

$$\begin{aligned} x &= w^H Y \\ x &= w^H HCA + w^H N \end{aligned} \quad (12)$$

The real part of  $x$  is sufficient and can be used to demodulate the user's bit

$$\Re(x) = \Re(w^H HCA) + \Re(w^H N) \quad (13)$$

Conditioned on the transmitted bit vector  $A$ ,  $\Re(x)$  is Gaussian with mean

$$\mu_x = \Re(w^H HCA) \quad (14)$$

and variance

$$\sigma_x^2 = \frac{\sigma^2 \|w\|^2}{2} \quad (15)$$

Since the demodulated bit is  $\text{sgn}(\Re(x))$ ,  $\text{sgn}(x)$  being the signum function, an error in demodulation occurs whenever

$$\Re(x|a_0 = -1) > 0 \text{ or } \Re(x|a_0 = +1) < 0 \quad (16)$$

Since  $\Re(x)$  is Gaussian, with mean  $\mu_x$  and variance  $\sigma_x^2$ , the pdf of  $\Re(x)$  is

$$f_{\Re(x|A)}(u) = \frac{1}{\sigma_x \sqrt{2\pi}} \exp\left(-\frac{(u - \mu_x)^2}{2\sigma_x^2}\right) \quad (17)$$

Conditioned on  $a_0 = +1$  the probability of error in demodulation is

$$P_e(x|a_0 = +1) = \int_{-\infty}^0 \frac{1}{\sigma_x \sqrt{2\pi}} \exp\left(-\frac{(u - \mu_x)^2}{2\sigma_x^2}\right) du \quad (18)$$

and conditioned on  $a_0 = -1$ , the probability of error in demodulation is

$$P_e(x|a_0 = -1) = \int_0^{\infty} \frac{1}{\sigma_x \sqrt{2\pi}} \exp\left(-\frac{(u - \mu_x)^2}{2\sigma_x^2}\right) du \quad (19)$$

Using the  $Q$  function, the probability of error can be written as

$$P_e(x|A) = Q\left(\frac{a_0 \mu_x}{\sigma_x}\right) \quad (20)$$

Since the  $2^N$  transmitted bit vectors are independent and equiprobable, the probability of error in demodulating the user can be written as

$$P_e = \frac{1}{2^N} \sum_{\forall A} Q\left(\frac{a_0 \mu_x}{\sigma_x}\right) \quad (21)$$

The filter vector that minimizes the probability of error is the MPOE detector.

#### B. MPOE Algorithm

The filter vector that minimizes the probability of error given by expression (21) can be found adaptively by two different approaches. The first approach requires computation of the gradient of the function to be minimized. The gradient may be computed analytically or numerically. Algorithms like conjugate gradient method, variable metric method fall in this category. The other approach requires function and not gradient evaluation. Algorithms that fall in this approach are downhill simplex method and Powell's method.

We have used the method of gradient descent, as the gradient of (21) can be easily calculated. Starting from an initial guess, the filter weights are updated iteratively along the negative gradient of the probability of error surface.

$$w_{k+1} = w_k - \mu \nabla (P_e(w)) \quad (22)$$

where  $\mu$  is a suitably chosen step size. To compute the gradient of (21) we note that if  $J(w)$  is a real valued cost function, then the gradient  $\nabla J$  is given by [9]

$$\nabla J = 2 \frac{\partial J}{\partial w^*} \quad (23)$$

$\frac{\partial J}{\partial w^*}$  is the conjugate derivative of the function  $J(w)$ . The conjugate derivative of (21) is given by

$$\frac{\partial P_e(w)}{\partial w^*} = -\frac{1}{\sqrt{2\pi}2^N} \sum_{\forall A} \left( e^{-\frac{\mu_x^2}{2\sigma_x^2}} \frac{\partial}{\partial w^*} \left( \frac{\mu_x a_0}{\sigma_x} \right) \right) \quad (24)$$

The term  $\frac{\partial}{\partial w^*} \left( \frac{\mu_x a_0}{\sigma_x} \right)$  can be written as

$$\begin{aligned} \frac{\partial}{\partial w^*} \left( \frac{\mu_x a_0}{\sigma_x} \right) &= \frac{\sigma_x a_0 \frac{\partial}{\partial w^*} \mu_x - \mu_x a_0 \frac{\partial}{\partial w^*} \sigma_x}{\sigma_x^2} \\ &= \frac{a_0}{\sigma_x} \left( \frac{\partial}{\partial w^*} \Re(w^H HCA) - \frac{\mu_x}{\sigma_x} \frac{\partial}{\partial w^*} \sigma \|w\| \right) \\ &= \frac{a_0}{\sigma_x} \left( HCA - \frac{\mu_x}{\|w\|} \frac{\partial}{\partial w^*} (w^H w)^{1/2} \right) \\ &= \frac{a_0}{\sigma_x} \left( HCA - \frac{\mu_x w}{\|w\|^2} \right) \end{aligned} \quad (25)$$

Using the above, the gradient of (21) can be written as

$$\frac{\partial P_e(w)}{\partial w^*} = -\frac{1}{\sqrt{2\pi}2^T} \sum_{\forall A} \left( a_0 e^{-\frac{\mu_x^2}{2\sigma_x^2}} \phi \right) \quad (26)$$

where  $\phi$  is

Fig. 1. BER of MPOE v/s MMSE for decreasing SNR for 8 users

Fig. 2. BER of MPOE v/s MMSE for increasing users at 10 dB SNR

$$\phi = \frac{1}{\sigma_x} \left( HCA - \frac{\mu_x w}{\|w\|^2} \right) \quad (27)$$

$w$  can be normalized to unit length for computational advantage, but that makes the algorithm more sensitive to the step size  $\mu$ . The computational complexity of the above algorithm can be reduced by noting that the expression (21) for the probability of error is a sum of  $2^N$  terms, one for each bit vector  $A$ . The sum remains unaffected if  $A$  is replaced by  $-A$ , i.e., all  $+1s$  are changed to  $-1s$  and vice versa. Therefore we need to sum up only  $2^{N-1}$  terms now instead of  $2^N$  terms. Complexity of the detector can be further reduced by using the Minimum Conditional Probability of Error (MCPOE) approach [7] [8].

#### IV. SIMULATIONS

Two sets of simulations were carried out to test the theoretical results. In the first set, the effect of noise on the performance of the MPOE detector was studied for a full load system. In the second set, the effect of increasing the load on the system was studied.

##### A. Simulation Parameters

The number of sub-carriers in the simulations were 8 and the number of maximum active users were also 8. Orthogonal Walsh-Hadamard codes were used to spread the incoming data stream. All the user's signals were assumed to be transmitted with the same power. Flat Rayleigh fading was assumed and also the fading was assumed to be independent for all the sub-carriers. Estimate of the channel matrix  $H$  was assumed to be known at the receiver. Also synchronization was assumed.

##### B. Results

Fig. 1 shows the effect of increasing the noise on the system. It can be seen that the MPOE detector outperforms the MMSE detector at all SNR levels.

Fig. 2 shows the effect of increasing the load on the system. The SNR is  $10dB$ . It can be seen that as the load on the system increases, the MPOE detector performs better than the MMSE detector.

#### V. CONCLUSIONS

A MPOE detector has been proposed in this paper for MC-CDMA. Performance of this detector has been compared with the MMSE detector. The MPOE detector attains a lower BER compared to the MMSE detector albeit at a slightly higher computational complexity. The computational complexity of the MPOE algorithm can be decreased by using the MCPOE approach [7] [8].

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