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(Article begins on next page)

# An ML optimal CDMA multiuser receiver

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*Indexing terms: Synchronous CDMA, Multiuser detection, Maximum likelihood detection,  
Voronoi regions*

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A maximum likelihood optimum detector with the asymptotic complexity per user  $O(1.5^K)$  has been derived for the synchronous DS/CDMA channel. The detector employs a local descent algorithm through the Voronoi regions for the equivalent hypothesis detector.

*Introduction:* In the last decade there has been much work concerning multiuser detection for Direct-Sequence Code Division Multiple Access (DS/CDMA) systems. The maximum likelihood (ML) optimum receiver for a synchronous CDMA system is a  $K$ -ary hypothesis detector, where  $K$  is the number of users in the system. The drawback though is that the asymptotic complexity per user for this receiver is  $O(2^K)$ . Thus much attention has been directed to suboptimal detectors [1]. Many of these multiuser receivers are sensitive to near-far effects, that is, when the received power from one user is much lower than the received power from another user. For example, the conventional receiver will only be near-far resistant if orthogonal codes are used. However, it has been shown that the decorrelator and multistage receivers are near-far resistant and that the ML optimum detector is optimal in this sense too [2, 3]. Hence, an ML optimum low complexity receiver would be very attractive. In this paper we present an ML optimum receiver with asymptotic complexity  $O(1.5^K)$ .

*System model:* Assume a synchronous DS/CDMA system with  $K$  users. User  $k$  transmit at a given time with power  $w_k$  the BPSK modulated bit  $b_k \in \{\pm 1\}$ . Then it can be shown that the matched filter outputs can be written as  $\mathbf{y} = \mathbf{R}\mathbf{W}\mathbf{b} + \mathbf{z}$ ,

where  $\mathbf{b} = (b_1, \dots, b_K)^T$  and  $\mathbf{W} = \text{diag}(w_1, \dots, w_K)$  [1]. The crosscorrelation matrix  $\mathbf{R}$  consists of the periodical (even) crosscorrelations between all codes of the users. The vector  $\mathbf{z}$  is additive coloured Gaussian noise with zero mean and covariance matrix  $\sigma^2\mathbf{R}$ . To decide on  $\mathbf{b}$  given an observed vector  $\mathbf{y}$ , the Maximum Likelihood criterion is usually expressed as

$$\hat{\mathbf{b}} = \arg \max_{\mathbf{b} \in \{-1, 1\}^K} \{2\mathbf{y}^T \mathbf{W} \mathbf{b} - \mathbf{b}^T \mathbf{W} \mathbf{R} \mathbf{W} \mathbf{b}\} \quad (1)$$

To whiten the noise (assuming  $\mathbf{R}$  to be positive definite), we transform the matched filter outputs  $\mathbf{y}$  into  $\mathbf{u} = \mathbf{R}^{-1/2} \mathbf{y} = \mathbf{T} \mathbf{b} + \mathbf{n}$ , where  $\mathbf{T} = \mathbf{R}^{1/2} \mathbf{W}$  and the covariance of the Gaussian noise  $\mathbf{n}$  is  $\sigma^2 \mathbf{I}$ .

If vectors are regarded as points in  $K$ -dimensional space, then the  $2^K$  vectors  $\mathbf{b}$  constitute the vertices of a hypercube. Similarly, the constellation

$$\mathcal{X} = \{\mathbf{T} \mathbf{b}\}_{\mathbf{b} \in \{-1, 1\}^K}$$

spans a  $K$ -dimensional parallelepiped, that is, a linear transform of a hypercube, where  $\mathbf{T}$  is the transformation matrix. Given an observation vector  $\mathbf{u}$ , the most likely point in  $\mathcal{X}$  is given by the minimum Euclidean distance to  $\mathbf{u}$ , i.e.

$$\hat{\mathbf{b}} = \arg \min_{\mathbf{b} \in \{-1, 1\}^K} \|\mathbf{u} - \mathbf{T} \mathbf{b}\|$$

This detection rule, which is equivalent to (1), partitions space into  $2^K$  convex polytopes, called *Voronoi regions*,

$$\mathcal{V}(\mathbf{x}) = \{\mathbf{u} \in \mathbb{R}^K : \|\mathbf{u} - \mathbf{x}\| \leq \|\mathbf{u} - \mathbf{x}'\|; \forall \mathbf{x}' \in \mathcal{X}\} \quad (2)$$

one for each  $\mathbf{x} \in \mathcal{X}$ . The ML decision on  $\mathbf{b}$  given  $\mathbf{u}$  is thus to find  $\hat{\mathbf{b}}$  such that  $\mathbf{u} \in \mathcal{V}(\mathbf{T} \hat{\mathbf{b}})$ .

*The new maximum likelihood receiver:* We employ an iterative approach, called *neighbour descent* (ND), to locate the Voronoi region that the observation belongs to [4]. It utilises the fact that out of the  $2^K - 1$  nontrivial inequalities in (2), some may be redundant, so that the same region can be described as

$$\mathcal{V}(\mathbf{x}) = \{\mathbf{u} \in \mathbb{R}^K : \|\mathbf{u} - \mathbf{x}\| \leq \|\mathbf{u} - \mathbf{x}'\|; \forall \mathbf{x}' \in \mathcal{N}(\mathbf{x})\} \quad (3)$$

The minimal set  $\mathcal{N}(\mathbf{x})$  for which this is true is called the set of *neighbours* of  $\mathbf{x}$ . Geometrically, each element of  $\mathcal{N}(\mathbf{x})$  corresponds to one facet ( $K-1$ -

dimensional face) of the polytope  $\mathcal{V}(\mathbf{x})$ . Thus, to check if a given vector  $\mathbf{u}$  belongs to the Voronoi region  $\mathcal{V}(\mathbf{x})$ , it is sufficient to compute  $|\mathcal{N}(\mathbf{x})|$  distances, a number that is often considerably smaller than  $2^K - 1$ . We compute these distances sequentially, and if a point  $\mathbf{x}' \in \mathcal{N}(\mathbf{x})$  is found for which  $\|\mathbf{u} - \mathbf{x}'\| < \|\mathbf{u} - \mathbf{x}\|$ , we immediately terminate the examination of  $\mathcal{N}(\mathbf{x})$ , replace  $\mathbf{x}$  with  $\mathbf{x}'$  as the best found vector, and restart. After a finite number of steps, the algorithm terminates at a vector  $\mathbf{x} = \mathbf{T}\mathbf{b}$  with the property that none of its neighbours is better. That  $\mathbf{u} \in \mathcal{V}(\mathbf{x})$  now follows from (3), and (2) completes the proof that  $\mathbf{x}$  is indeed the global optimum. Thus, the ML decoded bit vector is  $\mathbf{b}$ .

The algorithm assumes that the neighbours  $\mathcal{N}(\mathbf{x})$  for all  $\mathbf{x} \in \mathcal{X}$  have been precomputed and stored in memory. To identify the neighbour pairs, the following statement is useful.

***Theorem:** If every pair of points in  $\mathcal{X}$  is joined by a line, the neighbours are given by the lines that are only intersected by longer lines.*

This theorem can be proved using the theory of *indecomposable vectors* by Verdú [5].

The lines between points in  $\mathcal{X}$  can be regarded as diagonals in a parallelepiped. They intersect each other at the points

$$\mathcal{S} = \{\mathbf{T}\mathbf{b}\}_{\mathbf{b} \in \{-1,0,1\}^K} \setminus \mathcal{X}$$

and nowhere else. Since at most one pair of neighbours can meet at each intersection point, the total number of neighbour pairs in the point set is upperbounded by  $|\mathcal{S}| = 3^K - 2^K$ . In other words, an average point in  $\mathcal{X}$  has fewer than  $2(3/2)^K$  neighbours. This number sets the complexity for the ND receiver, since most of the time is used for verification, that is, comparing the final point  $\mathbf{x}$  to its neighbours [4].

In any parallelepiped, there are several congruent faces. Because the lines passing through any point in  $\mathcal{S}$  have the same lengths as the lines through one of the points in the subset

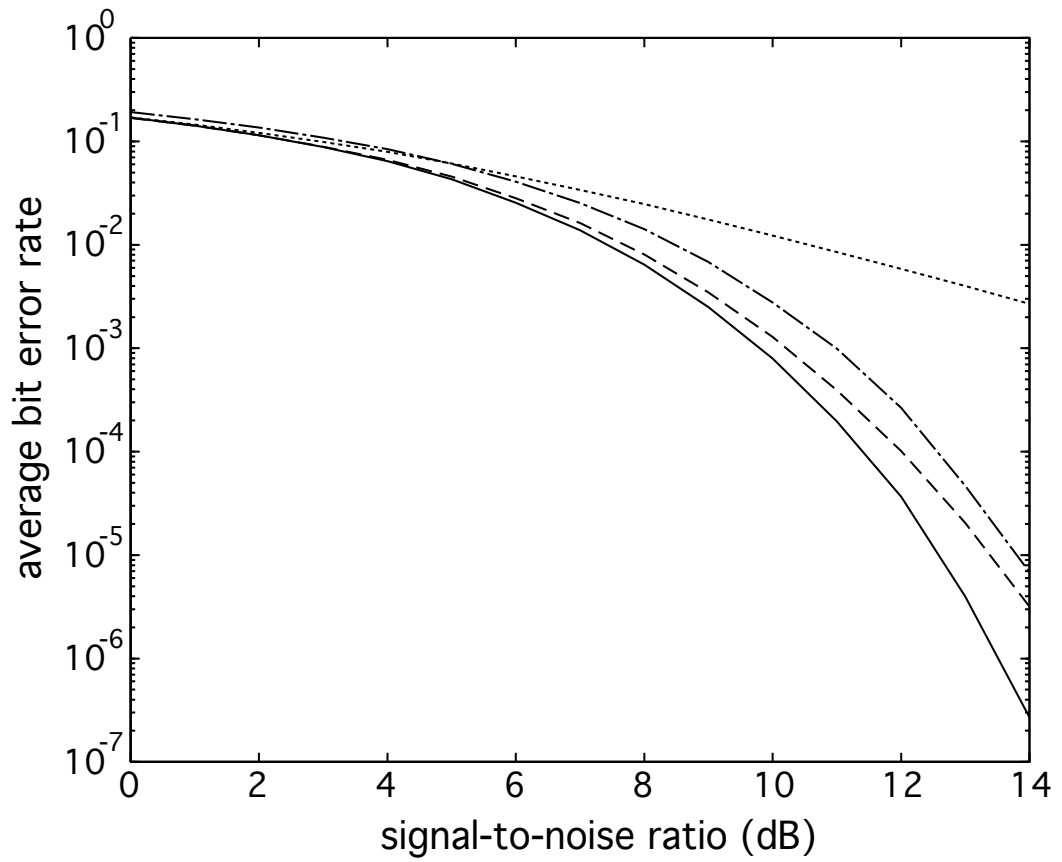
$$\mathcal{S}' = \{\mathbf{T}\mathbf{b}\}_{\mathbf{b} \in \{0,1\}^K \setminus \{1\}^K}$$

a full list of neighbours is obtained by examining these  $2^K - 1$  intersection points. The same property can be used to reduce the amount of memory needed for storage of the neighbours [6].

*Numerical results:* The ND algorithm described above was tested on a system using Gold-sequences of period 7. For comparison, we also simulated the conventional detector (matched filters followed by signum decisions), the decorrelator [1], and the multistage detector with two stages and the conventional detector as the first stage [3]. The average bit error rate for these detectors versus the signal-to-noise ratio for  $K = 6$  users is shown in Fig. 1. Considerable performance gain is achieved at high signal-to-noise ratios. For a system with either more users or worse correlation properties, the performance gain is even greater. Further, the performance gain for the ML receiver is, as stated above, higher in near-far situations.

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- ..... conventional
- · - · decorrelator
- - - multistage
- ND

**Fig. 1** Bit error rate versus signal-to-noise ratio for  $K = 6$  users.