

ELE 533 Multiuser Communication Theory

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Selected Problems: 4.12, 4.20, 4.58

1 Problem 4.12

Solution: According to the construction in the book, we 'convert' the asynchronous CDMA model to equivalent synchronous CDMA model, and we will compute $\Omega(\mathbf{b})$, which is the payoff function:

$$\begin{aligned}\mathbf{b}^T \mathbf{H} \mathbf{b} &= \sum_{j=1-MK}^{MK+K} \sum_{l=1-MK}^{MK+K} b_j b_l h_{j,l} \\ &= \sum_{j=1-MK}^{MK+K} b_j [A_{\kappa(j)}^2 b_j + 2 \sum_{l=j+1}^{j+K-1} b_l h_{l,j}] \\ &= \sum_{j=1-MK}^{MK+K} b_j [A_{\kappa(j)}^2 b_j + 2 \sum_{n=1}^{K-1} b_{j+n} h_{j+n,j}] \\ &= \sum_{j=1-MK}^{MK+K} b_j [A_{\kappa(j)}^2 b_j + 2 \sum_{n=1}^{K-1} b_{j+n} A_{\kappa(j)} A_{\kappa(j+n)} \rho_{\kappa(j), \kappa(j+n)}] \quad (1)\end{aligned}$$

This comes from the property of symmetric matrix: $\forall \mathbf{Q} \in \mathbb{R}^{N \times N}$ and $\mathbf{x} \in \mathbb{R}^{N \times 1}$, we have:

$$\begin{aligned}
\mathbf{x}^T \mathbf{Q} \mathbf{x} &= \sum_{i=1}^N \sum_{j=1}^N Q_{ij} x_i x_j \\
&= \sum_{i=1}^N [Q_{ii} x_i^2 + \sum_{j=1, j \neq i}^N Q_{ij} x_i x_j] \\
&= \sum_{i=1}^N [Q_{ii} x_i^2 + \sum_{j=1}^{i-1} Q_{ij} x_i x_j + \sum_{j=i+1}^N Q_{ij} x_i x_j] \tag{2}
\end{aligned}$$

We further note that:

$$\begin{aligned}
\sum_{i=1}^N \sum_{j=1}^{i-1} Q_{ij} x_i x_j &= \sum_{j=1}^{N-1} \sum_{i=j+1}^N Q_{ij} x_i x_j \\
&= \sum_{j=1}^N \sum_{i=j+1}^N Q_{ij} x_i x_j \\
&= \sum_{i=1}^N \sum_{j=i+1}^N Q_{ji} x_i x_j \quad \text{by interchanging } i \text{ and } j \\
&= \sum_{i=1}^N \sum_{j=i+1}^N Q_{ij} x_i x_j \quad \text{by the symmetry of } Q \tag{3}
\end{aligned}$$

Plug (3) into (2) we have:

$$\mathbf{x}^T \mathbf{Q} \mathbf{x} = \sum_{i=1}^N [Q_{ii} x_i^2 + 2 \sum_{j=i+1}^N Q_{ij} x_i x_j] \tag{4}$$

From equation (4) we can get equation (1)

Then we can compute $\Omega(\mathbf{b})$:

$$\begin{aligned}
\Omega(\mathbf{b}) &= 2\mathbf{b}^T \mathbf{A}_M \mathbf{y} - \mathbf{b}^T \mathbf{H} \mathbf{b} \\
&= 2 \sum_{j=1-MK}^{MK+K} A_{\kappa(j)} b_j y_j - \sum_{j=1-MK}^{MK+K} b_j [A_{\kappa(j)}^2 b_j + 2 \sum_{n=1}^{K-1} b_{j+n} A_{\kappa(j)} A_{\kappa(j+n)} \rho_{\kappa(j), \kappa(j+n)}] \\
&= \sum_{j=1-MK}^{MK+K} A_{\kappa(j)} b_j [2y_j - A_{\kappa(j)} b_j - 2 \sum_{n=1}^{K-1} b_{j+n} A_{\kappa(j+n)} \rho_{\kappa(j), \kappa(j+n)}] \\
&= \sum_{j=1-MK}^{MK+K} \mu_j(b_j, \mathbf{x}_j) \tag{5}
\end{aligned}$$

where

$$\mu_j(b_j, \mathbf{x}_j) = A_{\kappa(j)} b_j [2y_j - A_{\kappa(j)} b_j - 2 \sum_{n=1}^{K-1} x_j(n) A_{\kappa(j+n)} \rho_{\kappa(j), \kappa(j+n)}] \quad (6)$$

$$\begin{aligned} \mathbf{x}_j &= (x_j(1), x_j(2), \dots, x_j(K-1))^T \\ &= (b_{j+1}, b_{j+2}, \dots, b_{j+K-1})^T \end{aligned} \quad (7)$$

$$\begin{aligned} \mathbf{x}_{j-1} &= (b_j, b_{j+1}, \dots, b_{j+K-2})^T \\ &= f(b_j, \mathbf{x}_j) \end{aligned} \quad (8)$$

$$\mathbf{x}_{\mathbf{MK}+\mathbf{K}} = \mathbf{0} \quad (9)$$

To maximize $\Omega(\mathbf{b})$ is equivalent to maximize $\tilde{\Omega}(\mathbf{b}) = \sum_{j=1-MK}^{MK+K} \tilde{\mu}_j(b_j, \mathbf{x}_j)$

,where:

$$\tilde{\mu}_j(b_j, \mathbf{x}_j) = A_{\kappa(j)} b_j [y_j - \sum_{n=1}^{K-1} x_j(n) A_{\kappa(j+n)} \rho_{\kappa(j), \kappa(j+n)}] \quad (10)$$

Then we can write down backward dynamic programming algorithm as follows:

$$J_{MK+K}(\mathbf{x}_{\mathbf{MK}+\mathbf{K}}) = 0 \quad (11)$$

$$\begin{aligned} J_{MK+K-1}(\mathbf{x}_{\mathbf{MK}+\mathbf{K}-1}) &= \max \{ \tilde{\mu}_{MK+K}(1, \mathbf{x}_{\mathbf{MK}+\mathbf{K}}), \\ &\quad \tilde{\mu}_{MK+K}(-1, \mathbf{x}_{\mathbf{MK}+\mathbf{K}}) \} \end{aligned} \quad (12)$$

$$\begin{aligned} J_{j-1}(\mathbf{x}_{j-1}) &= \max \{ J_j(x_{j-1}(2), x_{j-1}(3), \dots, x_{j-1}(K-1), 1) + \\ &\quad \tilde{\mu}_j(x_{j-1}(1), x_{j-1}(2), \dots, x_{j-1}(K-1), 1), \\ &\quad J_j(x_{j-1}(2), x_{j-1}(3), \dots, x_{j-1}(K-1), -1) + \\ &\quad \tilde{\mu}_j(x_{j-1}(1), x_{j-1}(2), \dots, x_{j-1}(K-1), -1) \} \end{aligned} \quad (13)$$

As the same in the forward dynamic programming algorithm, since each level we need to calculate the J_j 's for 2^{K-1} states, so the time complexity per bit is $O(2^K)$

2 Problem 4.20

Solution: We first condition on b_2 , so this becomes a simple binary hypothesis testing problem with the following hypotheses:

$$H1 : y(t) = (1 + A_2 b_2)^2 s^2(t) + \sigma n(t) \quad (14)$$

$$H2 : y(t) = A_2^2 b_2^2 s^2(t) + \sigma n(t) \quad (15)$$