Homework Set #5Due 17 November 2009, 6 pm, in INR036

## **Problem 1** (There are almost no perfect codes)

Let  $\mathcal{C}$  be a linear binary *perfect code* consisting of binary sequences of length N. Assume that for the rate of code C we have  $R_C > 0$  where  $R_C \triangleq \frac{\log_2 |\vec{C}|}{N}$ . In this problem we would like to show that useful perfect codes do not exist (here, "useful"

means having large block-length N, and rate close neither to 0 nor 1).

Let  $\alpha \in (1/3, 1/2)$  be a parameter. In this problem we will show that there is no large perfect code that is  $\alpha N$ -error-correcting.

Remember that a code is *perfect*  $\alpha N$ -*error-correcting code* if the set of  $\alpha N$ -spheres centered on the codewords of the code fill the Hamming space without overlapping.

Let us suppose that such a code has been found.

- (a) Knowing that the code is  $\alpha N$ -error-correcting code, what can we say about its minimum distance?
- (b) Let us focus just on three codewords of this code. (Remember that the code has rate  $R_{\mathcal{C}} > 0$ , so it should have  $2^{NR_{\mathcal{C}}}$  codewords which is a large number if N grows.) Without loss of generality, we choose one of the codewords to be the all-zero codeword and define the other two to have overlaps with it as shown in the following

$c_0 = 000000$	00000000000000	000000	0000
$c_1 = 111111$	111111111111111	000000	0000
$c_2 = \underbrace{000000}$	1111111111111	<u>111111</u>	$\underbrace{0000}$
uN	vN	wN	xN

where u + v + w + x = 1.

Use the distance property of code  $\mathcal{C}$  to show that it cannot even have three codewords  $c_0$ ,  $c_1$ , and  $c_2$  (let alone  $2^{NR_c}$  codewords).

## Problem 2 (Reed-Solomon Codes)

- (a) Show that if H is the parity check matrix of a code of length n, then the code has minimum distance at least d if every d-1 columns of H are linearly independent.
- (b) Consider a linear code defined over a finite field  $\mathbb{F}$  with the parity check matrix

$$H = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ \alpha_1 & \alpha_2 & \cdots & \alpha_n \\ \alpha_1^2 & \alpha_2^2 & \cdots & \alpha_n^2 \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_1^{n-k-1} & \alpha_2^{n-k-1} & \cdots & \alpha_n^{n-k-1} \end{bmatrix}_{(n-k) \times n}$$

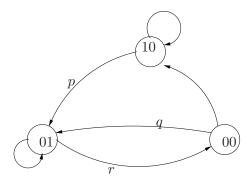


Figure 1: Problem 3

where  $k \leq n \leq |\mathbb{F}|$  and  $\alpha_i \in \mathbb{F}$  such that  $\alpha_i \neq \alpha_j$  if  $i \neq j$ . A matrix with this form called a *Vandermonde matrix*. It can be shown that the parity check matrix of a Reed-Solomon code is in fact a Vandermonde matrix.

Show that every n - k columns of H are linearly independent. Hint: For a square  $n \times n$  Vandermonde matrix

$$V = \begin{bmatrix} 1 & 1 & \cdots & 1 \\ \alpha_1 & \alpha_2 & \cdots & \alpha_n \\ \alpha_1^2 & \alpha_2^2 & \cdots & \alpha_n^2 \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_1^{n-1} & \alpha_2^{n-1} & \cdots & \alpha_n^{n-1} \end{bmatrix}_{n \times n}$$

we have

$$\det(V) = \prod_{1 \le i < j \le n} (\alpha_j - \alpha_i).$$

(c) From part (b) and the Singelton bound conclude that the Reed-Solomon codes are maximum distance seperable codes.

## Problem 3

We have a source that produces a sequence of bits with the following two properties:

- A "1" is always followed by a "0",
- No more than three "0"s come in a row.

Assume that this source can be modeled by a first order Markov chain as shown in Fig 1

- (a) Choose p, q, and r such that the entropy rate of this Markov process is maximized.
- (b) Construct a 2-state FSM that receives the source outputs as its input and maximally compresses it.
- (c) Is this finite state machine uniquely decodable?
- (d) Is this finite state machine information lossless?

## Problem 4 (LEMPEL-ZIV ALGORITHM IS ASYMPTOTICALLY OPTIMAL)

Consider a first order Markov process  $X_0, X_1, \cdots$  with the stationary distribution  $[p_0, p_1, \cdots, p_m]$ , where  $p_i$  denotes the stationary distribution of being in state  $i \in \{0, \cdots, m\}$ . Assume that the Markov process is in state 0. We define  $T_0$  as the number of steps it takes for the process to return to state 0 again.

- (a) Calculate  $\mathbb{E}T_0$  for a 2-state Markov process in terms of  $p_0$  and  $p_1$ .
- (b) Define  $s_i$  as the expected number of visits to state *i* before returning from 0 to state 0. i.e.,

$$s_i = \mathbb{E}_0[\sum_{n \ge 1} 1_{\{X_n = i\}} 1_{\{n \le T_0\}}],$$

where the index 0 of  $\mathbb{E}_0$  shows the fact that we are considering the chain from the time it has left state 0. Show that

$$p_i = \frac{s_i}{\sum_j s_j}$$

and conclude that  $p_0 = \frac{1}{\mathbb{E}(T_0)}$ .

(c) Take the Markov process  $X_0, X_1, \cdots$  and form the following extended Markov process from it:  $X_0^{n-1}, X_1^n, X_2^{n+1}, \cdots$ . How many steps does it take on average for this extended process to return for the first time to the state  $00 \cdots 0$  (after it left it).

In the LZ77 algorithm with infinite-length sliding window, in order to encode the block  $x_0x_1 \cdots x_{n-1}$ , one finds and communicates the last time the *n* symbols have been seen. Call it  $R_n(x_0x_1 \cdots x_{n-1})$ . If we denote the length of description of  $R_n(X_0X_1 \cdots X_{n-1})$  by  $l(X_0X_1 \cdots X_{n-1})$ , it can easily be shown that

$$\lim_{n \to \infty} \frac{1}{n} \mathbb{E}l(X_0 X_1 \cdots, X_{n-1}) = H(\mathcal{X})$$

and this is the basic idea of the proof of optimality of LZ77 algorithm. Refer to Homework 5 of last year's homeworks for details of proof.