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## Decoding Reed-Muller codes by multi-layer perceptrons

YUEN-HSIEN TSENG† and JA-LING WU†

The decoding of  $r$ th-order Reed-Muller codes of blocklength  $2^m$ , capable of correcting up to  $(1/2)2^{m-r} - 1$  errors, by use of multi-layer perceptrons is illustrated. The multi-layer perceptrons used have all their weights either  $+1$  or  $-1$  and all thresholds integers, thus making them very suitable for hardware implementation.

### 1. Introduction

Error-correcting codes, having wide applications in data transmission, data storage, and fault-tolerance computing, is to protect information from accidental errors. In this letter, we consider the decoding of a class of error-correcting codes called Reed-Muller codes. The Reed-Muller codes, used to transmit the Mariner photographs of Mars in 1972 (Blahut 1983), cover a wide range of rate and minimum distance. For each integer  $m$  and  $r < m$ , there is a Reed-Muller code of blocklength  $n = 2^m$  and information length,  $k$ ,

$$k = \binom{m}{0} + \binom{m}{1} + \cdots + \binom{m}{r}$$

capable of correcting up to  $(1/2)2^{m-r} - 1$  errors. The decoding rule of Reed-Muller codes involves only parity and majority operations, both of which can be implemented by multi-layer perceptrons, as is shown below.

A basic element in a multi-layer perceptron can be described as

$$z = \text{sgn}(g(X))$$

In the above equation, bipolar vector  $X = [x_1, x_2, \dots, x_n] \in \{+1, -1\}^n$  comes from the network input (or from the output of a previous layer);  $\text{sgn}$  is a sign function:  $\text{sgn}(a) = 1$  if  $a > 0$ ,  $-1$  if  $a \leq 0$ ; and  $g$  is a linear discriminant function:

$$g(X) = w_0 + w_1 x_1 + w_2 x_2 + \cdots + w_n x_n$$

where  $w_i$  are called weights and  $w_0$  is a threshold. The majority operation is to determine which one dominates the other in a sequence of  $+1$  and  $-1$ . If  $B = \{x_1, x_2, \dots, x_n\}$  denotes this sequence, then a single perceptron is able to perform the majority operation (Widrow and Winter 1988):

$$\text{maj}(x_1, x_2, \dots, x_n) = \text{sgn} \left( \sum_{i=1}^n x_i \right)$$

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The parity problem (or equivalent modulo-2 addition) is to determine if there is an odd number of 1 in a sequence of 0 and 1. A two-layer perceptron for the parity problem has been presented by Rumelhart *et al.* (1986). Here we reformulate the network such that it operates in bipolar domain and its weights are +1 or -1, thresholds are integers, and output unit is linear. Specifically, for a bipolar sequence  $B$ , define

$$S = n - \sum_{i=1}^n x_i$$

$S \in \{0, 2, 4, \dots, 2n\}$ . Let the parity function  $P(x_1, x_2, \dots, x_n)$  output -1 if there is an odd number of -1, and +1 otherwise. Then we have

$$P(x_1, x_2, \dots, x_n) = 1 - \sum_{k=1}^{\lfloor \frac{n+1}{2} \rfloor} [\text{sgn}(4k-1-S) + \text{sgn}(S-(4k-3))]$$

where the value in the square bracket is 2 if  $S=2(2k-1)$ , and 0 otherwise. Note this is a two-layer network with one hidden layer of approximately  $n$  units and one output unit which needs no sign function.

## 2. Decoding of Reed-Muller codes

A Reed-Muller code is best defined by a procedure for constructing a non-systematic generator matrix that will prove convenient for decoding. To construct the generator matrix, first define the vector product of two vectors by a component-wise multiplication. That is, let  $A=[a_1, a_2, \dots, a_n]$  and  $B=[b_1, b_2, \dots, b_n]$ , then  $AB=[a_1b_1, a_2b_2, \dots, a_nb_n]$ . The generator matrix for the  $r$ th-order  $(n, k)$  Reed-Muller code is defined as an array of blocks

$$G = \begin{bmatrix} G_0 \\ G_1 \\ \vdots \\ G_r \end{bmatrix}$$

where  $G_0$  is the vector of length  $2^m$  containing all ones;  $G_1$ , an  $m$  by  $2^m$  matrix, has each binary representation of  $j$  ranging from 0 to  $2^m-1$  appearing once as a column, with zero in the leftmost column and low-order bits in the bottom row; and  $G_r$  is constructed from  $G_1$  by taking its rows to be all possible vector products of any  $r$  rows of  $G_1$ .

An example is the zeroth-order  $(n, 1)$  Reed-Muller code. It is a simple repetition code and is decoded trivially by a majority vote. As another example, let  $m=4$ ,  $n=16$ , and  $r=3$ . Then

$$G_0 = [1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1 \ 1] = [B_{-1}]$$

$$G_1 = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} B_3 \\ B_2 \\ B_1 \\ B_0 \end{bmatrix}$$

$G_2$  has  $\binom{4}{2}$  rows since  $G_1$  has four rows,

$$G_2 = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} B_3 B_2 \\ B_3 B_1 \\ B_3 B_0 \\ B_2 B_1 \\ B_2 B_0 \\ B_1 B_0 \end{bmatrix}$$

and  $G_3$  has  $\binom{4}{3}$  rows

$$G_3 = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} B_3 B_2 B_1 \\ B_3 B_2 B_0 \\ B_3 B_1 B_0 \\ B_2 B_1 B_0 \end{bmatrix}$$

The generator matrix for this third-order (16, 15) Reed-Muller code is

$$G = \begin{bmatrix} G_0 \\ G_1 \\ G_2 \\ G_3 \end{bmatrix}$$

In fact, it is simply a parity-check code. Note we have given each row in the above a label to indicate its relation with the rows of  $G_1$  for later convenience.

The Reed algorithm is designed specially for decoding Reed-Muller codes (Blahut 1985). The Reed algorithm is unusual as compared to most decoding algorithms for most codes in that it recovers the information bits directly from the received word without computing the syndrome vectors. The basic idea is to construct a decoder for an  $r$ th-order Reed-Muller code by reducing it to a decoder for an  $(r-1)$ th-order Reed-Muller code. Because the 0th-order Reed-Muller code can be decoded by majority vote, the complete decoder is then established by induction. A more detail description of this idea is given below.

Let the information vector be broken into  $r+1$  segments, written  $A = [A_0, A_1, \dots, A_r]$ , where segment  $A_i$  contains  $\binom{m}{i}$  information bits. Each segment multiplies one block of  $G$ . The encoding can be represented as a block vector-matrix product

$$C = AG \text{ mod } 2 = [A_0, \dots, A_r] \begin{bmatrix} G_0 \\ \vdots \\ G_r \end{bmatrix} \text{ mod } 2$$

The information bits are distributed into the codeword  $C$ . The received word is  $V = [v_0, v_1, \dots, v_{n-1}] = C + E$ , where  $E$  is an error pattern. If we can recover  $A_r$  from  $V$ , then we can compute their contribution to the received word and subtract this contribution. That is

$$V' = V - A_r G_r \text{ mod } 2 = [A_0, \dots, A_{r-1}] \begin{bmatrix} G_0 \\ \vdots \\ G_{r1} \end{bmatrix} + E \text{ mod } 2$$

This reduces the problem to that of decoding an  $(r-1)$ th-order Reed-Muller code.

As to recovering  $A_r$  from  $V$ , consider decoding the information bit  $a_j$  which multiplies one row of  $G_r$ . This is decoded by setting up  $2^{m-r}$  linear check sums in the  $2^m$  received bits; each such check sum involves  $2^r$  bits of the received word, and each received bit is used in only one check sum. The check sums will be formed so that  $a_j$  contributes to only one term of each check sum, and every other information bit contributes to an even number of terms in each check sum. Hence, each check sum is equal to  $a_j$  in the absence of errors. But if there are at most  $(1/2)2^{m-r} - 1$  errors, the majority of the check sums will still equal  $a_j$ .

Specifically, if  $a_j$  corresponds to the row of  $G_r$  labelled  $B_{i_1} B_{i_2} \dots B_{i_r}$ , one of the  $2^{m-r}$  check sums starting with  $v_k$ , derived from the description given by Wosley Patterson and Weldon (1972), is expressed as follows.

$$\begin{aligned} \hat{a}_j(s) = & v_k + \left. \begin{matrix} \\ \\ \\ \end{matrix} \right\} \binom{r}{0} \text{ term} \\ & v_{k+2^{i_1}} + v_{k+2^{i_2}} + \dots + v_{k+2^{i_r}} + \left. \begin{matrix} \\ \\ \\ \end{matrix} \right\} \binom{r}{1} \text{ terms} \\ & v_{k+2^{i_1}+2^{i_2}} + \dots + v_{k+2^{i_{r-1}}+2^{i_r}} + \left. \begin{matrix} \\ \\ \\ \end{matrix} \right\} \binom{r}{2} \text{ terms} \\ & + \dots + \dots \\ & v_{k+2^{i_1}+2^{i_2}+\dots+2^{i_r}} \left. \begin{matrix} \\ \\ \\ \end{matrix} \right\} \binom{r}{r} \text{ term} \\ & \text{mod } 2 \end{aligned}$$

where  $s$  ranges from 1 to  $2^{m-r}$ . The information bit  $a_j$  is obtained by taking the majority of these  $\hat{a}_j(1)$ . The received bit  $v_j$  without the contribution of the information bits in segment  $A_r$  can then be computed by

$$v'_j = v_j - \sum_{i=r'}^r a_i g_{ij} \text{ mod } 2$$

where  $g_{ij} \in \{1, 0\}$  is an entry of  $G$ , and

$$r' = \binom{m}{0} + \binom{m}{1} + \cdots + \binom{m}{r-1}$$

$$r'' = r' + \binom{m}{r} - 1$$

Because subtraction in modulo 2 is equivalent to addition,  $v'_j$  is rewritten as

$$v'_j = v_j + \sum_{i=r'}^{r''} a_i g_{ij} \pmod{2}$$

Since the above decoding algorithm involves only addition and majority operations over binary elements, decoding of Reed-Muller codes by multi-layer perceptrons is now in order.

### Theorem

An  $r$ th-order Reed-Muller code can be decoded by a  $2(r+1)$ -layer perceptron. The Reed-Muller decoder is implemented by the foregoing parity networks followed by the majority networks. Since the output of the parity network is linear, it can be immediately directed to the input of the majority networks. This makes it a two-layer perceptron for decoding each information segment.

### Example

The first-order (8,4) Reed-Muller code, or equivalently the single-error correcting/double error detecting Hamming code, can be decoded by a multi-layer perceptron with four outputs:

$$z_3 = \text{sgn}(P(x_0, x_1) + P(x_2, x_3) + P(x_4, x_5) + P(x_6, x_7))$$

$$z_2 = \text{sgn}(P(x_0, x_2) + P(x_1, x_3) + P(x_4, x_6) + P(x_5, x_7))$$

$$z_1 = \text{sgn}(P(x_0, x_4) + P(x_1, x_5) + P(x_2, x_6) + P(x_3, x_7))$$

$$z_0 = \text{sgn}(x_0 + P(x_1, z_3) + P(x_2, z_2) + P(x_3, z_2, z_3) + P(x_4, z_1) + P(x_5, z_1, z_3) \\ + P(x_6, z_1, z_2) + P(x_7, z_2, z_3))$$

If there are two errors in the received word, there will be no majority in  $z_1$ ,  $z_2$ , or  $z_3$ , in which case a double-error pattern is detected.

### 3. Conclusions

We have illustrated a direct implementation of the Reed algorithm to decode  $r$ th-order Reed-Muller codes by  $2(r+1)$ -layer perceptrons. The used multi-layer perceptrons have all their weights at  $+1$  or  $-1$  and all thresholds integers, thus making them simple for hardware implementation.

The multi-layer perceptrons in this paper are constructed by concatenating two operations that are implementable by perceptrons. The idea of regarding some well-known networks as building blocks could be an alternative approach in constructing multi-layer perceptrons for solving problems.

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