

## A Syndrome-distribution decoding MOLS $L_p$ codes

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ABSTRACT. Let  $p$  be an odd prime number. We introduce simple and useful decoding algorithm for orthogonal Latin square codes of order  $p$ . Let  $H$  be the parity check matrix of orthogonal Latin square code. For any  $x \in GF(p)^n$ , we call  $xH^T$  the syndrome of  $x$ . This method is based on the syndrome decoding for linear codes. In  $L_p$ , we need to find the first and the second coordinates of codeword in order to correct the errored received vector.

### 1. Introduction

The organization of this paper is as follows: In Section 1, we will recall the well-known definitions concerning Latin squares and maximum set of orthogonal Latin squares. And we will summarize a construction of  $p - 1$  mutually orthogonal Latin squares when  $p$  is a prime number [7].

In Section 2, for an odd prime  $p$ , we will review a  $p$ -ary codes of specified minimum distance corresponding to  $p - 1$  mutually orthogonal Latin squares [4]. And using the weight enumerator of this code, we will find the minimum distance of its dual code. So we shall show how its dual codes is to Hamming codes of order 2 over  $GF(p)$ .

In 1970, D. C. Bossen, R. T. Chien and M. Y. Hsiao [2] have constructed a class of decodable multiple error-correcting codes which is based on one-step majority decoding method. In Section 3, we will prove the theorems which provide an algorithm for orthogonal Latin square codes in Section 4. Finally, we will give a syndrome-decoding algorithm and examples corresponding to each steps of this algorithm.

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**Definition 1.** A Latin square of order  $p$  is  $p \times p$  square array of numbers from an  $p$ -symbol alphabet (say  $0, 1, \dots, p-1$ ) in which each row and each column contains each symbol exactly once. Two Latin squares of the same order are (pairwise-)orthogonal if, when one Latin square is superimposed on the other, every ordered pair of elements are distinct. In particular, a set of Latin squares of the same order, any pair of which are orthogonal, is called a set of mutually (pairwise-)orthogonal Latin squares(MOLS).

Notice that we can permute rows and columns of the array preserving the Latin square property. So, we can always permute the rows and columns of the array so that the elements in the initial row and initial column are ordered. Also, if two Latin squares are orthogonal, the relabeling can be done independently for each square without destroying orthogonality.

To obtain a code corresponding to a set of mutually orthogonal Latin squares, it is important to determine the maximum possible number of mutually orthogonal Latin squares of given order  $p$ . Since [3], it is well known that  $p - 1$  is an upper bound. In particular if  $p$  is a prime number, there exist exactly  $p - 1$  mutually orthogonal Latin squares.

**Theorem 1 [3].** *For any  $p$ , there are at most  $p - 1$  mutually orthogonal Latin squares of order  $p$ .*

**Definition 2.** A set of  $p - 1$  mutually (pairwise-)orthogonal Latin squares of order  $p$  is said to be a complete of mutually (pairwise-)orthogonal Latin squares.

Let  $p$  be an odd prime. Then there exists a finite field  $GF(p)$  with  $p$  elements. Take an  $p \times p$  array

$$L_t = [u_t(i, j)], \quad 0 \leq i, j \leq p-1, \quad 1 \leq t \leq p-1$$

and in the cell  $(i, j)$  of this array put the integer  $u_t = u_t(i, j)$  given by

$$u_t = t \cdot i + j$$

where  $t$  is a fixed nonzero element of  $GF(p)$ . We write down the following Latin

square  $L_t$  of order  $p$ ,  $1 \leq t \leq p - 1$ ,

0	1	...	$p - 1$
$t$	$t + 1$	...	$t + p - 1$
$2t$	$2t + 1$	...	$2t + p - 1$
$\vdots$	$\vdots$	$\vdots$	$\vdots$
$(p-1)t$	$(p-1)t + 1$	...	$(p-1)t + p - 1$

where all expressions are to be taken mod  $p$ . In [1] and [7], we have seen that  $\{L_1, \dots, L_{p-1}\}$  is a set of  $p - 1$  orthogonal Latin squares.

As an example, we can write down a set of four orthogonal Latin squares of order 5,

		$L_1$						$L_2$			
0	1	2	3	4		0	1	2	3	4	
1	2	3	4	0		2	3	4	0	1	
2	3	4	0	1		4	0	1	2	3	
3	4	0	1	2		1	2	3	4	0	
4	0	1	2	3		3	4	0	1	2	
		$L_3$						$L_4$			
0	1	2	3	4		0	1	2	3	4	
3	4	0	1	2		4	0	1	2	3	
1	2	3	4	0		3	4	0	1	2	
4	0	1	2	3		2	3	4	0	1	
2	3	4	0	1		1	2	3	4	0	

In addition, when  $p$  is a prime power, we can get a similar result [7]. So we will not discuss them here.

## 2. Orthogonal Latin square codes and its dual codes

S. W. Golomb and E. C. Posner [4] established an important connection between the existence of sets of mutually orthogonal Latin squares and the existence of  $p$ -ary codes.

The following two concepts are equivalent:

- (1) A set of  $p - 1$  mutually orthogonal Latin squares of order  $p$ ,
- (2) An existence of linear code with length  $p + 1$ , minimum distance  $p$ ,  $p^2$  codeword.

The  $[p+1,2,p]$  code derived from  $p - 1$  mutually orthogonal Latin squares of order  $p$  is orthogonal Latin square codes of order  $p$ . From Section 1 and the above two concepts, we have the codewords as the form  $(i, j, i + j, \dots, (p - 1) \cdot i + j)$ ,  $0 \leq i, j \leq p - 1$ .

This construction has been generalized to multi-orthogonal higher dimensional Latin hypercubes by Silverman [8]. In his terms, an orthogonal Latin square code is equivalent to a set of  $d - 1$  mutually  $(n - d + 1)$ -wise orthogonal  $(n - d + 1)$ -dimensional Latin hypercubes where  $n, d$ , is the length and minimum distance respectively.

For any given two cells  $(i, j), (i', j')$ , we have the  $(t + 2)$ -th coordinate  $t \cdot i + j, t \cdot i' + j'$  of the codewords corresponding to  $(i, j), (i', j')$  respectively. Since  $(t \cdot i + j) + (t \cdot i' + j')$  is the  $(t + 2)$ -th coordinate of the codeword corresponding to  $(i + i', j + j')$ , a  $[p + 1, 2, p]$  orthogonal Latin square code is linear code with generator matrix G

$$\begin{bmatrix} 1 & 0 & 1 & 2 & 3 & \dots & (p - 1) \\ 0 & 1 & 1 & 1 & 1 & \dots & 1 \end{bmatrix} = [I_2 \ P],$$

where  $I_2$  is  $2 \times 2$  identity matrix and

$$P = \begin{bmatrix} 1 & 2 & 3 & \dots & (p - 1) \\ 1 & 1 & 1 & \dots & 1 \end{bmatrix}.$$

Hence the parity check matrix H of orthogonal Latin square code  $\mathcal{L}_p$  is :

$$H = [-P^T \ I_{p-1}] = \begin{bmatrix} p - 1 & p - 1 & 1 & 0 & \dots & 0 \\ p - 2 & p - 1 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & & & \dots & \\ 1 & p - 1 & 0 & 0 & \dots & 1 \end{bmatrix},$$

where  $I_{p-1}$  is  $(p - 1) \times (p - 1)$  identity matrix and  $P^T$  is transpose of P .

**Example 1.** Let  $p = 3$ . Then there exist 2 mutually orthogonal Latin squares,

	$L_1$				$L_2$		
	0	1	2		0	1	2
	1	2	0		2	0	1
	2	0	1		1	2	0

Thus we get a  $[4, 2, 3]$  orthogonal Latin square code  $\mathcal{L}_3$  of order 3 over  $\text{GF}(3)$  with 9 codewords

(0 0 0 0)	(0 1 1 1)	(0 2 2 2)
(1 0 1 2)	(1 1 2 0)	(1 2 0 1)
(2 0 2 1)	(2 1 0 2)	(2 2 1 0)

Then this code  $\mathcal{L}_3$  is the only self-dual linear orthogonal Latin square code because for a 2-dimensional linear code  $\mathcal{L}_p$ , the dimension of self-dual code  $\mathcal{L}_p^\perp$  is  $p - 1$  and  $p - 1 = 2$  only when  $p = 3$ .

From [9], it is easy to see that a  $[p + 1, 2, p]$  orthogonal Latin square code  $\mathcal{L}_p$  is a maximum distance separable (MDS) code. Thus this code has the maximum possible distance between codewords.

Consider the dual code  $\mathcal{L}_p^\perp$  of  $\mathcal{L}_p$ . It is well known that the error correcting capability of a code is determined by the minimum distance between all pairs of distinct codewords. Since  $\mathcal{L}_p$  is linear, so is  $\mathcal{L}_p^\perp$ . Thus the minimum distance of  $\mathcal{L}_p^\perp$  is equal to the minimum weight among all non-zero codewords of  $\mathcal{L}_p^\perp$ . So if we use the weight enumerator of  $\mathcal{L}_p^\perp$ , we can find not only the minimum distance but also the error correcting capability of  $\mathcal{L}_p^\perp$ . Let  $A_i$  denote the number of codewords of weight  $i$  in  $\mathcal{L}_p$ . The number of codewords of weight  $i$  in  $\mathcal{L}_p$  over  $\text{GF}(p)$  has been completely determined in [5] and [6], i.e.

$$\begin{aligned}
 A_i &= \binom{p+1}{i} \sum_{j=0}^{i-p} (-1)^j \binom{i}{j} (p^{i-p+1-j} - 1) \\
 &= \binom{p+1}{i} (p-1) \sum_{j=0}^{i-p} (-1)^j \binom{i-1}{j} p^{i-p-j}.
 \end{aligned}$$

Thus

$$\begin{aligned} A_p &= \binom{p+1}{p} (p-1) \\ &= p^2 - 1. \end{aligned}$$

So, the weight enumerator  $A(z)$  of  $\mathcal{L}_p$  is  $A_0 + A_p z^p$  where  $A_0 = 1$ ,  $A_p = p^2 - 1$ . Let  $B(z)$  denote the weight enumerator of the dual code  $\mathcal{L}_p^\perp$ . Then by [5] and [6] we have

$$\begin{aligned} B(z) &= p^{-2} (1 + (p-1)z)^{p+1} A\left(\frac{1-z}{1+(p-1)z}\right) \\ &= p^{-2} (1 + (p-1)z) \{(1 + (p-1)z)^p + (p^2 - 1)(1-z)^p\} \\ &= 1 + B_3 z^3 + \dots, \end{aligned}$$

where  $B_3$  is the number of codewords of weight 3. Thus the minimum distance of  $\mathcal{L}_p^\perp$  is 3, and so  $\mathcal{L}_p^\perp$  is a single-error-correcting code. By [5], the dual code  $\mathcal{L}_p^\perp$  is a  $[p+1, p-1, 3]$  MDS linear code over  $\text{GF}(p)$ . Therefore  $\mathcal{L}_p^\perp$  is perfect code because  $(1 + (p+1)(p-1)) \cdot p^{p-1} = p^{p+1}$  where  $1 + (p+1)(p-1)$  is the number of vectors in a sphere of radius 1 about a codeword and  $p^{p-1}$  is the number of spheres.

**Theorem 2 [9].** *Let  $C$  be an  $[n, k, d]$  code over  $\text{GF}(p)$  with parity check matrix  $H$ .  $C$  is MDS if and only if every  $n - k$  columns of  $H$  are linearly independent.*

From Theorem 2 and concept for the Hamming code of order 2, the dual code  $\mathcal{L}_p^\perp$  is the Hamming code of order 2 over  $\text{GF}(p)$ .

### 3. Main Theorems

In this section, all the arithmetic operations (i.e. addition and multiplication) are based on  $\text{GF}(p)$ .

For convenience, we first define the following notation:

$\mathbf{c} = (c_1, \dots, c_{p+1})$  : codeword in  $\mathcal{L}_p$ .

$\mathbf{r} = (r_1, \dots, r_{p+1})$  : received word.

$\mathbf{e} = (e_1, \dots, e_{p+1})$  : error vector.

i.e.  $\mathbf{r} = \mathbf{c} + \mathbf{e}$ .

$\mathbf{H}$  : parity check matrix (see previous Section).

$\mathbf{s} = (s_1, \dots, s_{p-1})$  : syndrome vector.

$\mathbf{s}(l) = \mathbf{s} - l \cdot (p-1, p-2, \dots, 2, 1) = (\hat{s}_1, \hat{s}_2, \dots, \hat{s}_{p-1})$  : dual syndrome  
with variable  $l$  for  $1 \leq l \leq p-1$ .

$M_b(\mathbf{s}) = \#\{i \mid s_i = b, 1 \leq i \leq p-1\}$  : syndrome distribution

for some syndrome  $\mathbf{s} = (s_1, \dots, s_{p-1})$  and some  $b \in \text{GF}(p)$ .

$M_b(\mathbf{s}(l)) = \#\{i \mid \hat{s}_i = b, 1 \leq i \leq p-1\}$  : dual syndrome distribution

for some dual syndrome  $\mathbf{s}(l)$  and some  $b \in \text{GF}(p)$ .

But, if codeword  $\mathbf{c}$  is changed into received word  $\mathbf{r}$  with error  $\mathbf{e}$ . Then  $\mathbf{s} = \mathbf{H}\mathbf{r}^T = \mathbf{H}(\mathbf{c} + \mathbf{e})^T = \mathbf{H}\mathbf{c}^T + \mathbf{H}\mathbf{e}^T = \mathbf{H}\mathbf{e}^T$ . So the  $i$ -th coordinate  $s_i$  of syndrome  $\mathbf{s}$  is  $s_i = -i \cdot e_1 - e_2 + e_{i+2}$ . Since  $\mathcal{L}_p$  has minimum distance  $p$ , we always assume that the Hamming weight of  $\mathbf{e}$  is less than or equal to  $\frac{p-1}{2}$ .

**Theorem 3.** Let  $\mathbf{r} = (r_1, \dots, r_{p+1})$  be a received word and  $\mathbf{s} = (s_1, \dots, s_{p-1})$  syndrome of  $\mathbf{r}$ .

(1) Both  $r_1$  and  $r_2$  are correct if and only if  $M_0(\mathbf{s}) \geq \frac{p-1}{2}$ .

(2)  $r_1$  is correct and  $r_2$  is not correct if and only if  $M_b(\mathbf{s}) \geq \frac{p+1}{2}$  for some  $b \in \text{GF}(p) - \{0\}$ .

*Proof of (1).* By previous paragraph,  $s_i = -i \cdot e_1 - e_2 + e_{i+2}$ ,  $1 \leq i \leq p-1$ .

( $\Rightarrow$ ) If both  $r_1$  and  $r_2$  are correct,  $e_1 = e_2 = 0$ . So,  $s_i \neq 0$  if and only if  $e_{i+2} \neq 0$ . But since Hamming weight of  $\mathbf{e}$  is less than or equal to  $\frac{p-1}{2}$ ,  $M_0(\mathbf{s}) \geq \frac{p-1}{2}$ .

( $\Leftarrow$ ) Suppose that  $r_1$  is correct and  $r_2$  is not correct (i.e.  $e_1 = 0$  and  $e_2 \neq 0$ ). Then  $s_i = 0$  if and only if  $e_2 = e_{i+2} \neq 0$ . But at most  $\frac{p-3}{2}$  elements of  $e_3, e_4, \dots, e_{p+1}$  are nonzero. i.e.  $M_0(\mathbf{s}) \leq \frac{p-3}{2}$ , which is contradict to hypothesis.

Suppose that  $r_1$  is not correct and  $r_2$  is correct (i.e.  $e_1 \neq 0$  and  $e_2 = 0$ ). Then, for  $i = 1, \dots, p-1$ ,  $s_i = 0$  if and only if  $i \cdot e_1 = e_{i+2}$ . But at most  $\frac{p-3}{2}$  elements of  $e_3, e_4, \dots, e_{p+1}$  are nonzero. i.e.  $M_0(\mathbf{s}) \leq \frac{p-3}{2}$ , which is contradict to hypothesis.

Suppose that both  $r_1$  and  $r_2$  are not correct (i.e.  $e_1 = e_2 \neq 0$ ). Then, for  $i = 1, \dots, p-1$ ,  $s_i = 0$  if and only if  $i \cdot e_1 + e_2 = e_{i+2}$ . But, for  $i = -\frac{e_2}{e_1}$ ,  $e_{i+2} = 0$  and for  $i \neq -\frac{e_2}{e_1}$ ,  $e_{i+2} \neq 0$ . But at most  $\frac{p-5}{2}$  elements of  $e_3, \dots, e_{p+1}$  are nonzero. i.e.  $M_0(\mathbf{s}) \leq 1 + \frac{p-5}{2} = \frac{p-3}{2}$ . This is contradict to hypothesis.

*Proof of (2).* ( $\Rightarrow$ ) By assumption,  $e_1 = 0$  and  $e_2 \neq 0$ . But since  $e_2 \neq 0$ , at least  $\frac{p+1}{2}$  elements of  $e_3, \dots, e_{p+1}$  are zero. So, for  $b = -e_2$ ,  $M_b(\mathbf{s}) \geq \frac{p+1}{2}$ .

( $\Leftarrow$ ) Suppose that  $r_1$  is not correct and  $r_2$  is correct (i.e.  $e_1 \neq 0$  and  $e_2 = 0$ ). But since  $e_1 \neq 0$ , at most  $\frac{p-3}{2}$  of  $e_3, \dots, e_{p+1}$  are nonzero. Hence, for  $b \neq 0$ ,  $\{i \mid s_i = -i \cdot e_1 + e_{i+2} = b\} \subset \{i \mid e_{i+2} = 0, i = -\frac{b}{e_1}\} \cup \{i \mid e_{i+2} \neq 0\}$ . Thus  $M_b(\mathbf{s}) \leq 1 + \frac{p-3}{2} = \frac{p-1}{2}$ . This is contradict to hypothesis.

Suppose that both  $r_1$  and  $r_2$  are not correct (i.e.  $e_1 \neq 0$ ,  $e_2 \neq 0$ ). Then at most  $\frac{p-5}{2}$  elements of  $e_3, \dots, e_{p+1}$  are nonzero. So, for  $b \neq 0$ ,  $\{i \mid s_i = -i \cdot e_1 - e_2 + e_{i+2} = b\} \subset \{i \mid e_{i+2} = 0\} \cup \{i \mid e_{i+2} \neq 0, i = -\frac{b+e_2-e_{i+2}}{e_1}\}$ . Hence  $M_b(\mathbf{s}) \leq 1 + \frac{p-5}{2} = \frac{p-3}{2}$ . This is contradict to hypothesis.  $\square$

**Theorem 4.** *Suppose that  $r_1$  is not correct (i.e. In Theorem 3, the conditions of (1) and (2) are not satisfied ).*

(1)  $r_2$  is correct if and only if  $M_0(\mathbf{s}(e_1)) \geq \frac{p+1}{2}$ .

(2)  $r_2$  is not correct if and only if for some  $b \neq 0$ ,  $M_b(\mathbf{s}(e_1)) \geq \frac{p+3}{2}$ .

*Proof of (1).* ( $\Rightarrow$ ) By definition, the  $i$ -th coordinate of dual syndrome  $\mathbf{s}(l)$  is  $\hat{s}_i = -i \cdot (e_1 - l) - e_2 + e_{i+2}$ . Hence by assumption  $e_2 = 0$  and at least for  $1 \leq i \leq p-1$  the number of  $e_{i+2}$  which is zero is greater than or equal to  $\frac{p+1}{2}$ . So  $M_0(\mathbf{s}(e_1)) \geq \frac{p+1}{2}$ .

( $\Leftarrow$ ) Suppose that  $r_2$  is not correct. Then for  $1 \leq i \leq p-1$ , the number of  $e_{i+2}$  which is not zero is less than or equal to  $\frac{p-5}{2}$ . So  $M_0(\mathbf{s}(e_1)) \leq \frac{p-5}{2}$ . This is contradict to hypothesis.



*Proof of (2).* ( $\Rightarrow$ ) By assumption, for  $1 \leq i \leq p-1$ , the number of  $e_{i+2}$  which is zero is greater than or equal to  $\frac{p+3}{2}$ . So  $b = -e_2$ ,  $M_b(s(e_1)) \geq \frac{p+3}{2}$ .

( $\Leftarrow$ ) Suppose that  $r_2$  is correct. Then for  $1 \leq i \leq p-1$ , the number of  $e_{i+2}$  which is not zero is less than or equal to  $\frac{p-3}{2}$ . So  $b \neq 0$ ,  $M_b(s(e_1)) \leq \frac{p-3}{2}$ . This is contradict to hypothesis.  $\square$

Note: since  $M_0(s(e_1))$  and  $M_b(s(e_1))$  (for some  $b \neq 0$ ) is greater than or equal to  $\frac{p+1}{2}$  in Theorem 4.  $M_0(s(e_1)) = M_b(s(e_1)) = \max_{b \in \text{GF}(p)} M_b(s(l))$  for  $1 \leq l \leq p-1$ . In (2) of Theorem 4, we can take  $i$  such that  $\hat{s}_i = b$ ,  $\hat{s}_{i+1} = b$ , because  $M_b(s(e_1)) \geq \frac{p+1}{2}$ . Then  $c_1 = r_{i+3} - r_{i+2}$ ,  $c_2 = r_{i+2} - i \cdot c_1$ .

#### 4. The syndrome-distribution method of $\mathcal{L}_p$ and examples

##### Algorithm

STEP 1 : If  $M_0(s) \geq \frac{p-1}{2}$ , then by Theorem 3-(1),  $r$  is decoded into  $c = (r_1, r_2, r_1 + r_2, \dots, (p-1)r_1 + r_2)$ .

STEP 2 : If  $M_0(s) < \frac{p-1}{2}$  and  $M_b(s) \geq \frac{p+1}{2}$  for  $b \neq 0$ , then by Theorem 3-(2),  $r$  is decoded into  $c = (r_1, B, r_1 + B, \dots, (p-1)r_1 + B)$  where  $B = r_2 + b$ .

STEP 3 : In the case that the conditions of Step 1 and Step 2 are not satisfied, if  $\max_{1 \leq l \leq p-1} M_0(s(l)) \geq \frac{p+1}{2}$ , then by Theorem 4-(1) codeword  $c = (A, r_2, A + r_2, \dots, (p-1)A + r_2)$ , where  $A = r_{i+3} - r_{i+2}$  for the  $i$  satisfying the condition  $\hat{s}_i = 0$ ,  $\hat{s}_{i+1} = 0$  (the  $i$ -th coordinate satisfying this statement exists, because  $\max_{1 \leq l \leq p-1} M_0(s(l)) \geq \frac{p+1}{2}$ ).

STEP 4 : In the case that the conditions of Step 1 and Step 2 are not satisfied, if for some  $b \neq 0$   $\max_{1 \leq l \leq p-1} M_b(s(l)) \geq \frac{p+3}{2}$ , then  $c = (A, B, A + B, \dots, (p-1)A + B)$ , where  $A = r_{i+3} - r_{i+2}$ ,  $B = r_{i+2} - i \cdot A$  for the  $i$  satisfying the condition  $\hat{s}_i = b$ ,  $\hat{s}_{i+1} = b$ .

**Example 2.** Let  $p = 5$ ,  $\mathbf{r} = (2, 3, 1, 3, 4, 1)$ .

$$\mathbf{H} = \begin{bmatrix} 4 & 4 & 1 & 0 & 0 & 0 \\ 3 & 4 & 0 & 1 & 0 & 0 \\ 2 & 4 & 0 & 0 & 1 & 0 \\ 1 & 4 & 0 & 0 & 0 & 1 \end{bmatrix}$$

is the parity check matrix for  $\mathcal{L}_5$  over  $\text{GF}(5)$ . Then the syndrome  $\mathbf{s}$  of  $\mathbf{r}$

$$\mathbf{H}\mathbf{r}^T = \begin{bmatrix} 4 & 4 & 1 & 0 & 0 & 0 \\ 3 & 4 & 0 & 1 & 0 & 0 \\ 2 & 4 & 0 & 0 & 1 & 0 \\ 1 & 4 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \\ 1 \\ 3 \\ 4 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}.$$

Since  $M_0(\mathbf{s}) \geq \frac{5-1}{2} = 2$ , both  $r_1$  and  $r_2$  are correct. By Step 1,  $\mathbf{c} = (2, 3, 2 + 3, 4 + 3, 1 + 3, 3 + 3) = (2, 3, 0, 2, 4, 1)$ .

**Example 3.** Let  $p = 5$ ,  $\mathbf{r} = (1, 3, 3, 1, 0, 1)$ . Then the syndrome  $\mathbf{s}$  of  $\mathbf{r}$  is  $(4, 1, 4, 4)$ . Since  $M_0(\mathbf{s}) < \frac{5-1}{2} = 2$  and  $M_4(\mathbf{s}) \geq \frac{5+1}{2} = 3$ , by Theorem 3-(2)  $r_1$  is correct. By Step 2, we have  $B = 3 + 4 = 2$  and  $\mathbf{c} = (1, 2, 3, 4, 0, 1)$ .

**Example 4.** Let  $p = 5$ ,  $\mathbf{r} = (3, 2, 1, 0, 2, 3)$ . Then the syndrome  $\mathbf{s}$  of  $\mathbf{r}$  is  $(1, 2, 1, 4)$ . From  $M_0(\mathbf{s}) < 2$ , for any  $b \neq 0$   $M_b(\mathbf{s}) < 3$  and  $\mathbf{s}(4) = (0, 0, 3, 0)$ , we get  $M_0(\mathbf{s}(4)) \geq 3$  and so by Step 3  $\mathbf{c} = (4, 2, 1, 0, 4, 3)$ .

**Example 5.** Let  $p = 5$ ,  $\mathbf{r} = (2, 1, 3, 4, 0, 1)$ . Then the syndrome  $\mathbf{s}$  of  $\mathbf{r}$  is  $(0, 4, 3, 2)$ . From  $M_0(\mathbf{s}) < 2$ , for any  $b \neq 0$   $M_b(\mathbf{s}) < 3$ , and  $\mathbf{s}(1) = (1, 1, 1, 1)$ , we get  $M_1(\mathbf{s}(1)) \geq 4$  and so by Step 4  $\mathbf{c} = (1, 2, 3, 4, 0, 1)$ .

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