

# Pixel Bit-Depth Increase by Bit Replication

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## ABSTRACT

In many applications, such as inverse dithering, it is necessary to increase the pixel bit-depth of images by expanding  $q$ -bit integer values to  $m$ -bit integer values ( $m > q$ ). This paper describes a simple and efficient method that uses bit replication, instead of conventional multiplication, to achieve this expansion. First, we show that the optimal number of repetitions is given by  $\text{ceiling}\{m/q\}$  and that the method is equivalent to multiplication by the ideal gain when  $m/q$  is an integer. We then demonstrate that, in the case where  $m/q$  is not an integer, truncating the fraction bits to the right of the decimal point will lead to zero average error. The paper also includes two suggestions for implementing the bit-replication process, both of which have a vast complexity advantage over a multiplier. Two examples are given at the end to illustrate the bit-replication process in action.

**Keywords:** computational precision, multiplication, hardware shortcut, inverse dithering

## 1. INTRODUCTION

There are several applications where the set of values within a particular range needs to be operated on at a larger, or higher-precision, range. An important example is an inverse-dithering system. When a digital image is dithered to a lesser number of displayable levels, the bit-depth of its pixel values is decreased. This has the effect of reducing the storage space required even though the perceptual quality of the image is largely preserved [1]. The goal of inverse dithering is to reconstruct from the dithered image the original image with a larger pixel bit-depth. Several approaches have been suggested, including adaptive algorithms [2][3], wavelet decomposition [4], and MAP estimation [5]. These approaches differ widely. However, when extended to multi-level-to-multi-level inverse dithering, they all require the essential step of expanding the pixel values in the dithered image from a lower bit-depth to their original bit-depth for subsequent processing.

In this bit-depth-increase step, a  $q$ -bit integer input level  $L_i$  with values from the range

$$L_i \in \{0, 1, 2, \dots, 2^q - 1\} \quad (1)$$

is scaled to a  $m$ -bit integer output level  $L_o$  with values from the range

$$L_o \in \{0, 1, 2, \dots, 2^m - 1\}. \quad (2)$$

This can be achieved by performing a multiplication by an *ideal gain*,

$$G = \frac{2^m - 1}{2^q - 1} \quad (3)$$

followed by rounding to the nearest integer:

$$L_o = \text{Round}\{G \times L_i\}. \quad (4)$$

Note that while a  $q$ -bit integer represents  $2^q$  integer values, it only represents  $2^q - 1$  intervals. Therefore the gain used in pixel-depth increase is not simply a case of left shifts followed by zero padding. In this paper we will show that multiplication by the ideal gain can be closely approximated by another simple implementation, that of bit replication. We will also determine the optimal number of bits to replicate for minimum error.

## 2. SERIES EXPANSION

For any real numbers  $a$  and  $b$ , where  $b$  is not equal to 0 or 1, the follow expression is valid:

$$\frac{a-1}{b-1} = \frac{a}{b} + \frac{(a/b)-1}{b-1}. \quad (5)$$

This expression can itself be used to expand the rightmost term further:

$$\frac{(a/b)-1}{b-1} = \frac{a}{b^2} + \frac{(a/b^2)-1}{b-1}. \quad (6)$$

Recursively repeating this expansion  $N$  times results in the series:

$$\frac{a-1}{b-1} = \sum_{i=1}^N \frac{a}{b^i} + \left( \frac{(a/b^N)-1}{b-1} \right). \quad (7)$$

In our case, using  $a = 2^m$  and  $b = 2^q$ , the ideal gain can be expressed as

$$G = \sum_{i=1}^N 2^{m-iq} + \left( \frac{2^{m-Nq} - 1}{2^q - 1} \right). \quad (8)$$

From this, we define the approximate gain to be

$$\hat{G} = \sum_{i=1}^N 2^{m-iq}, \quad (9)$$

and the gain-error to be

$$E = \frac{1 - 2^{m-Nq}}{2^q - 1}. \quad (10)$$

For the purpose of analysis, let us assume that the output level  $L_o$  can be a real number (instead of an integer) and the rounding operation is unnecessary. The data scaling will therefore be implemented as follows:

$$L_o = \hat{G} \times L_i. \quad (11)$$

Under these assumptions, the gain-error  $E$  is proportional to the error in the output level.

## 3. BIT REPLICATION

We can show that multiplication by the approximate gain  $\hat{G}$  can be implemented simply as a replication of the input level bits. Since each term in  $\hat{G}$  is a power of 2, multiplication is equivalent to a shift of the input. Note also that the shifts

corresponding to successive terms differ by  $q$  spaces, which is the same as the number of bits in the input. This means that there is no overlap of bits and no addition is necessary. Consider the example of converting a 3-bit input ( $q = 3$ ) to an 8-bit output ( $m = 8$ ). In Figure 1 the values of the 3 input bits are “A”, “B”, and “C” as shown. If the sum was carried out to 5 terms or repeat periods ( $N = 5$ ), the output value would have the bit values as indicated. This bit replication extends into the fraction bits to the right of the decimal point.

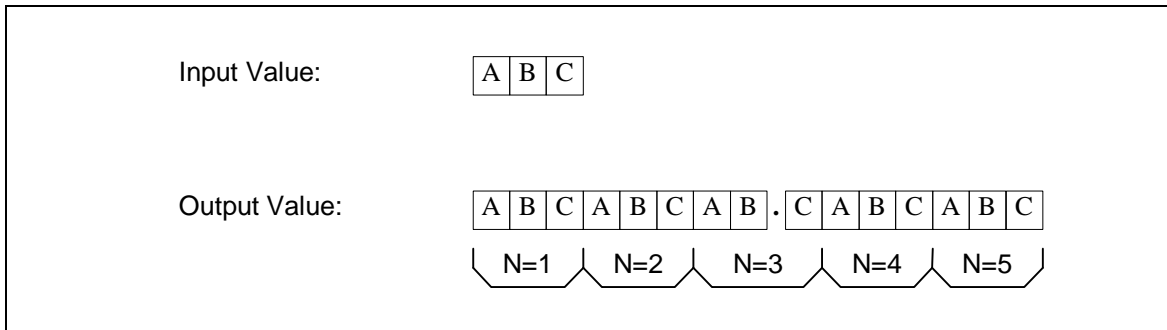


Figure 1. Example of scaling a 3-bit input value to 8 bits using the approximate gain  $\hat{G}$  with  $N = 5$ .

The weighting of the bit values in this representation is shown in Figure 2. Multiplying an input by the approximate gain will result in a real number with an integer part and a fraction part. Although the goal is to generate a  $m$ -bit integer output value, it may be useful to keep additional fraction bits for intermediate computations of the output value for higher accuracy.

In the next sections two aspects of the bit-replication process will be optimized in terms of minimizing the error. First,

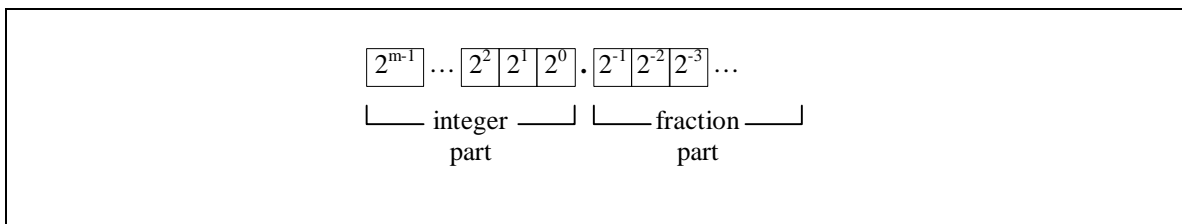


Figure 2. Representation of a real number with binary digits.

the optimal value of  $N$  is determined. Then for cases where the last repeat periods have both integer and fraction bits (as in the third period of Figure 1) the contribution of the fraction bits to the accuracy of the output level is found.

#### 4. OPTIMIZATION OF REPEAT PERIODS

The optimal number of repeat periods  $N$  can be determined simply by solving for the value of  $N$  that minimizes  $|E|$ .  $|E|$  is zero exactly once when  $m - Nq = 0$ , so the optimal value should be

$$N = m / q . \tag{12}$$

The problem is that  $N$  can only take on integer values so this result will only occur when  $q$  is an exact multiple of  $m$ . However, if this condition does hold, the error is identically zero, and bit replication yields a perfectly scaled output level.

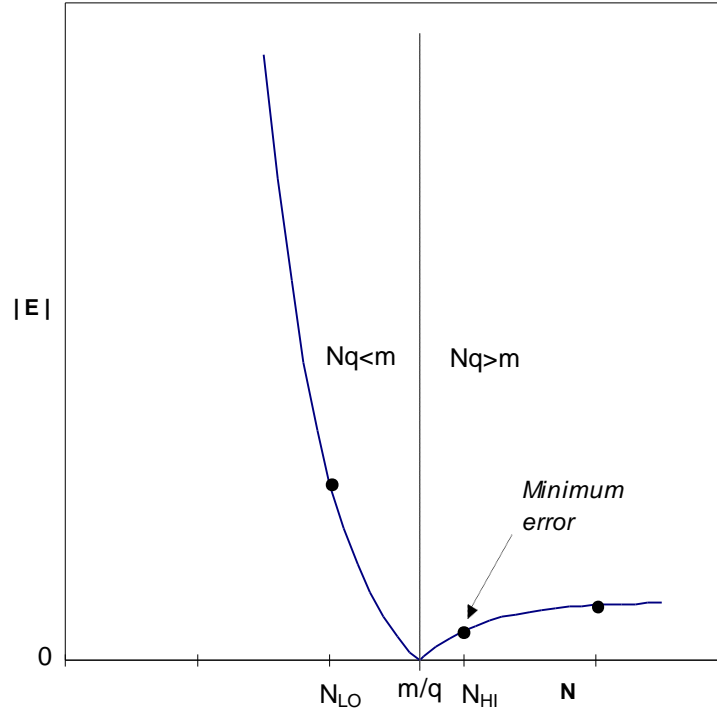


Figure 3. Absolute error as a function of  $N$ .

For the case where  $(m/q)$  is not an integer, the solution for the optimum  $N$  is more involved. In Figure 3  $|E|$  is plotted as a function of  $N$ . The graph reveals two candidate integer values for optimum  $N$  on either side of the point  $(m/q)$ :  $N_{LO}$  is the nearest integer less than  $(m/q)$  and  $N_{HI}$  is the nearest integer greater than  $(m/q)$ . Since the denominator of  $|E|$  is constant with respect to  $N$ , we can consider the equivalent problem of minimizing the function

$$\mathbf{b} = \left| 2^{m-Nq} - 1 \right|. \quad (13)$$

For  $N < m/q$ ,  $\mathbf{b}$  increases monotonically with decreasing  $N$  because the exponent  $(m - Nq)$  becomes increasingly positive. This exponent is a positive integer, so the *smallest*  $\mathbf{b}$  can be is 1.

For  $N > m/q$ ,  $\mathbf{b}$  increases monotonically with increasing  $N$  because the exponent  $(m - Nq)$  becomes increasingly negative. ( $|E|$  asymptotically approaches  $1/(2^q - 1)$ .) The exponent  $(m - Nq)$  is a negative integer, so  $\mathbf{b} < 1$ .

Therefore, of the two candidates,  $N_{LO}$  and  $N_{HI}$ , the value of  $N$  that minimizes  $|E|$  is  $N_{HI}$ . This is the nearest integer larger than the non-integer value  $(m/q)$ , or equivalently,

$$N = \text{ceiling}\{m/q\}. \quad (14)$$

## 5. CONTRIBUTION OF FRACTION BITS

Consider the case of  $q = 5$  and  $m = 8$ . The optimal number of repeat periods is

$$N = \text{ceiling}\{8/5\} = 2 \quad (15)$$

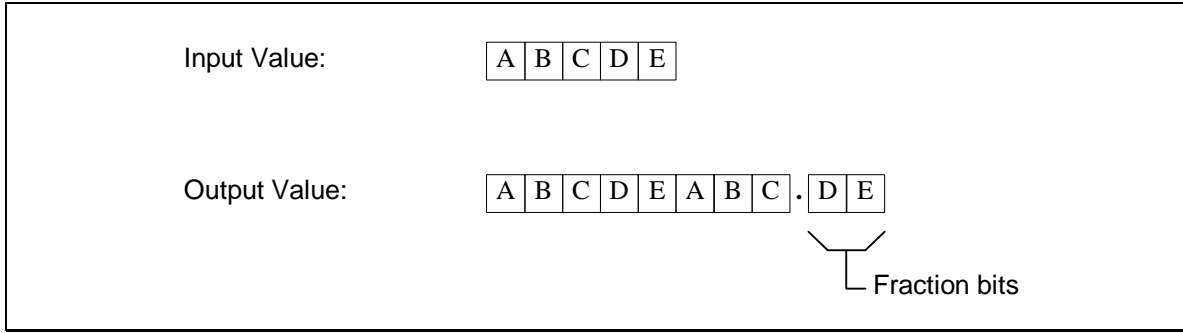


Figure 4. Scaling a 5-bit number to an 8-bit number with  $N = 2$  repeat periods.

For a 5-bit input, the output is as shown in Figure 4. There are 2 fraction bits for this particular example. In all cases, the number of fraction bits due to repeating bits by the optimum  $N$  periods is

$$b = Nq - m . \quad (16)$$

Note that there are no fraction bits for the error-free case where  $(m/q)$  is an integer.

In the following, we derive the average contribution of the fraction bits over all possible input levels. Since the fraction bits are simply the  $b$  least significant bits of each input level  $L_i$ , they will take on the values  $\{0,1,2,\dots,2^b - 1\}$  in a periodic fashion as  $L_i$  increases. Note that there will be  $2^{q-b}$  complete periods in the range of  $L_i$ . The average contribution of the fraction bits over all  $L_i$  is therefore the same as the average contribution over one single period.

Let the value of the fraction bits be  $i$  and the contribution of the fraction bits to the output be  $F$ . We can then write

$$F = \frac{i}{2^b} \quad (17)$$

and the average contribution of the fraction bits to the output is therefore given by

$$\bar{F} = \frac{1}{2^b} \sum_{i=0}^{2^b-1} \frac{i}{2^b} . \quad (18)$$

Using the fact that

$$\sum_{i=0}^{M-1} i = M \left( \frac{M-1}{2} \right), \quad (19)$$

the above expression simplifies to

$$\bar{F} = \frac{1-2^{-b}}{2} . \quad (20)$$

Next, we derive the average output error due to the gain-error over all possible input levels. Using the change of variables  $b = Nq - m$ , we can rewrite the gain-error to be

$$E = \frac{1-2^{-b}}{2^q - 1} . \quad (21)$$

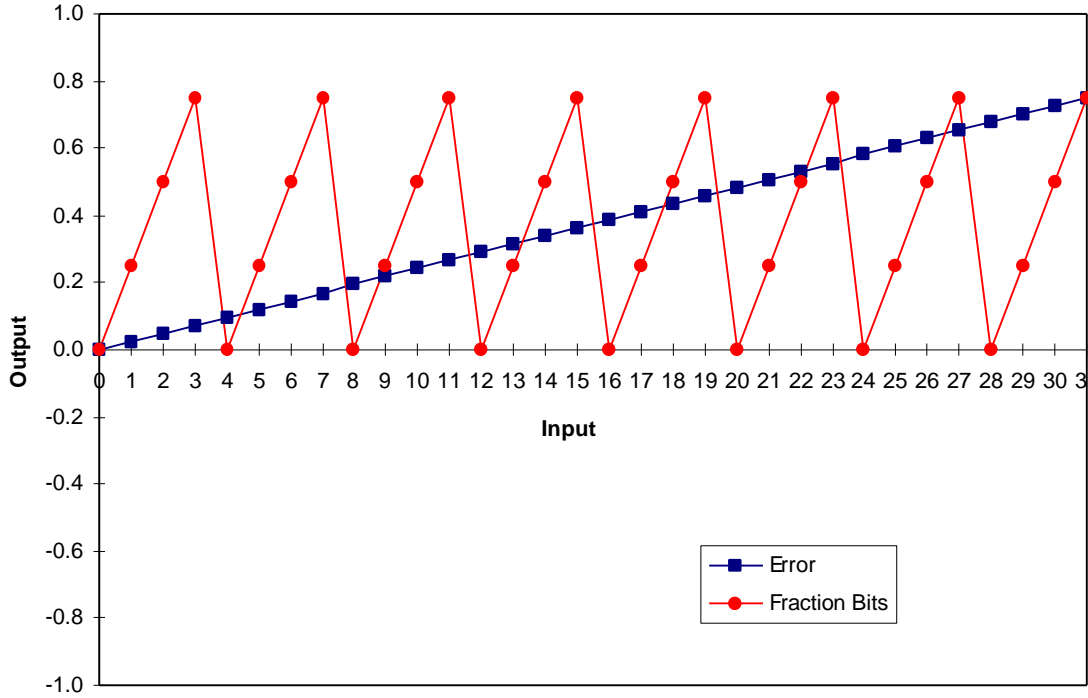


Figure 5. Output error as a function of input, and the contribution of fraction bits to the output value.

For each input level  $L_i$ , the output error  $e$  is  $E \times L_i$ . The average output error is then given by

$$\bar{e} = \frac{1}{2^q} \sum_{L_i=0}^{2^q-1} L_i \left( \frac{1-2^{-b}}{2^q-1} \right) = \frac{1-2^{-b}}{2}. \quad (22)$$

For our example of  $q=5$  and  $m=8$ , the contribution of the fraction bits  $F$  and output error  $e$  are plotted in Figure 5.

It is very interesting to note that for all cases,

$$\bar{e} = \bar{F}. \quad (23)$$

By this we can conclude that zero average error can be achieved by always truncating the fraction bits! Using our example of  $q=5$  and  $m=8$  again, the output error is plotted in Figure 6 for the cases of no fraction bits used, one fraction bit used, and both fraction bits used. The average errors are 0.000, 0.25, and 0.375 ( $= (1-2^{-2})/2$ ) respectively.

## 6. IMPLEMENTATION

We have proved that multiplication of a  $q$ -bit input level  $L_i$  by the ideal gain

$$G = \frac{2^m - 1}{2^q - 1} \quad (24)$$

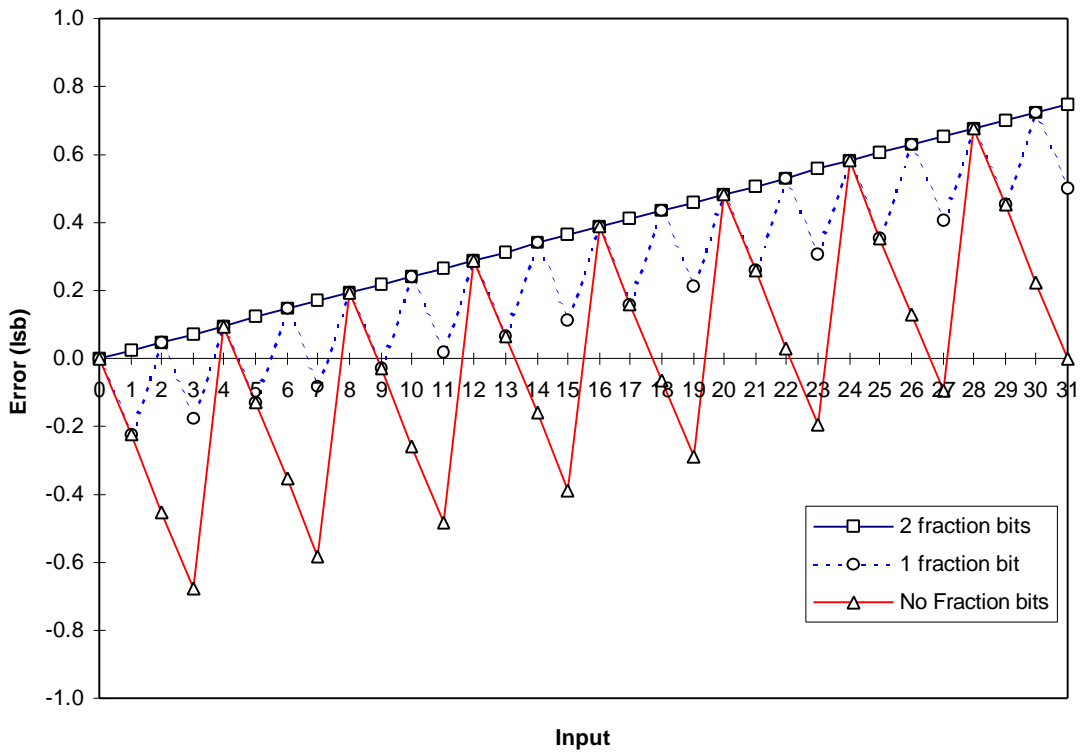


Figure 6. Output error as a function of number of fraction bits retained.

can be achieved with *zero average error* by simply replicating the input bits from the most significant part of the  $m$ -bit output register to the least significant part. Using additional fraction bits for intermediate computation can only increase average error.

The most straightforward hardware implementation of this bit replication method is the wired approach as illustrated in Figure 7. In this figure we use the case of  $q = 3$  and  $m = 8$  as an example. “A”, “B”, and “C” represent input bit values.

Another implementation can use shift registers as depicted in Figure 8. The input register performs a circular shift left into an output shift register as shown. The operation is complete after  $m$  shifts.

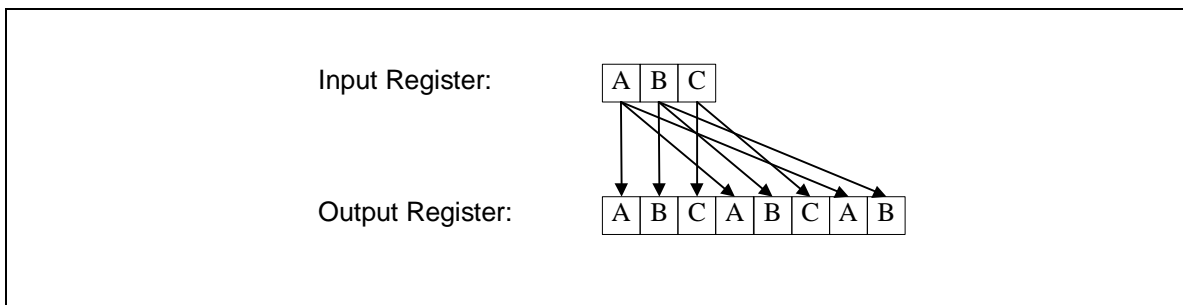


Figure 7. Wired implementation.

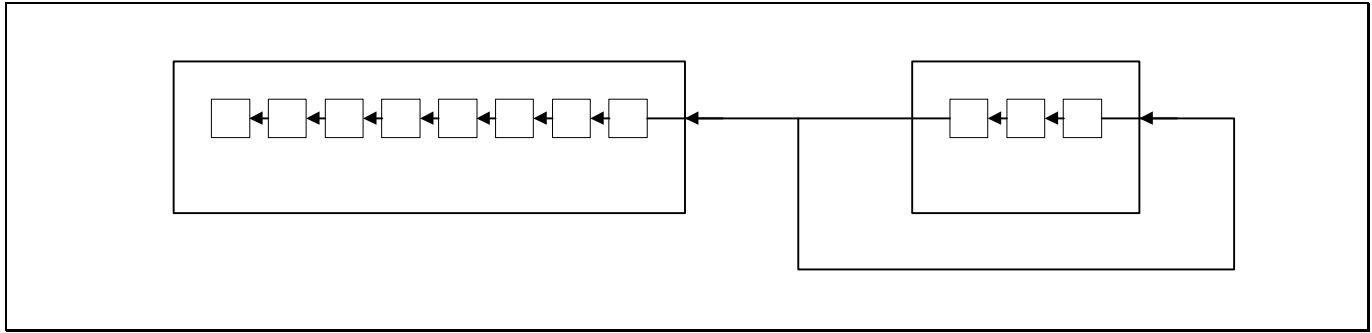


Figure 8. Shift-register implementation

Either case has a tremendous complexity advantage over a multiplier.

## 7. EXAMPLES

We will end this article by using two examples to illustrate the bit-replication process in action. The first example is the case of  $q = 5$  and  $m = 8$ . In Table 1, we show, for each possible input level, the actual output level and its difference from the ideal output level. Note that the average error over all possible input levels is zero. Table 2 shows the case of  $q = 4$  and  $m = 12$ . In this case,  $m/q$  is an integer, so the bit-replication gives the exact answer and the error is zero for all input levels.

## 8. REFERENCES

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Input $L_i$		Output $L_o$		Ideal	Error
00000	( 0 )	00000000	( 0 )	0.00	0.00
00001	( 1 )	00001000	( 8 )	8.23	-0.23
00010	( 2 )	00010000	( 16 )	16.45	-0.45
00011	( 3 )	00011000	( 24 )	24.68	-0.68
00100	( 4 )	00100001	( 33 )	32.90	0.10
00101	( 5 )	00101001	( 41 )	41.13	-0.13
00110	( 6 )	00110001	( 49 )	49.35	-0.35
00111	( 7 )	00111001	( 57 )	57.58	-0.58
01000	( 8 )	01000010	( 66 )	65.81	0.19
01001	( 9 )	01001010	( 74 )	74.03	-0.03
01010	( 10 )	01010010	( 82 )	82.26	-0.26
01011	( 11 )	01011010	( 90 )	90.48	-0.48
01100	( 12 )	01100011	( 99 )	98.71	0.29
01101	( 13 )	01101011	( 107 )	106.94	0.06
01110	( 14 )	01110011	( 115 )	115.16	-0.16
01111	( 15 )	01111011	( 123 )	123.39	-0.39
10000	( 16 )	10000100	( 132 )	131.61	0.39
10001	( 17 )	10001100	( 140 )	139.84	0.16
10010	( 18 )	10010100	( 148 )	148.06	-0.06
10011	( 19 )	10011100	( 156 )	156.29	-0.29
10100	( 20 )	10100101	( 165 )	164.52	0.48
10101	( 21 )	10101101	( 173 )	172.74	0.26
10110	( 22 )	10110101	( 181 )	180.97	0.03
10111	( 23 )	10111101	( 189 )	189.19	-0.19
11000	( 24 )	11000110	( 198 )	197.42	0.58
11001	( 25 )	11001110	( 206 )	205.65	0.35
11010	( 26 )	11010110	( 214 )	213.87	0.13
11011	( 27 )	11011110	( 222 )	222.10	-0.10
11100	( 28 )	11100111	( 231 )	230.32	0.68
11101	( 29 )	11101111	( 239 )	238.55	0.45
11110	( 30 )	11110111	( 247 )	246.77	0.23
11111	( 31 )	11111111	( 255 )	255.00	0.00
<b>average error</b>					0.00

Table 1: The Input Levels, the Output Levels, and the Errors for the Case of  $q = 5$  and  $m = 8$

Input $L_i$		Output $L_o$		Ideal	Error
0000	( 0 )	000000000000	( 0 )	0.00	0.00
0001	( 1 )	000100010001	( 273 )	273.00	0.00
0010	( 2 )	001000100010	( 546 )	546.00	0.00
0011	( 3 )	001100110011	( 819 )	819.00	0.00
0100	( 4 )	010001000100	( 1092 )	1092.00	0.00
0101	( 5 )	010101010101	( 1365 )	1365.00	0.00
0110	( 6 )	011001100110	( 1638 )	1638.00	0.00
0111	( 7 )	011101110111	( 1911 )	1911.00	0.00
1000	( 8 )	100010001000	( 2184 )	2184.00	0.00
1001	( 9 )	100110011001	( 2457 )	2457.00	0.00
1010	( 10 )	101010101010	( 2730 )	2730.00	0.00
1011	( 11 )	101110111011	( 3003 )	3003.00	0.00
1100	( 12 )	110011001100	( 3276 )	3276.00	0.00
1101	( 13 )	110111011101	( 3549 )	3549.00	0.00
1110	( 14 )	111011101110	( 3822 )	3822.00	0.00
1111	( 15 )	111111111111	( 4095 )	4095.00	0.00
<b>average error</b>					0.00

Table 2: The Input Levels, the Output Levels, and the Errors for the Case of  $q=4$  and  $m=12$ .