

Improving JPEG-LS Performance Using Location Information

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Abstract

JPEG-LS is an international standard for lossless or near-lossless image-compression algorithms. In this paper, a simple method is proposed to improve the performance of the lossless JPEG-LS algorithm. With respect to JPEG-LS and its supplementary explanation, Golomb-Rice (GR) coding is mainly used for entropy coding, but it is not used for long codewords. The proposed method replaces a set of long codewords with a set of shorter location map information. This paper shows how efficiently the location map guarantees reversibility and enhances the compression rate in terms of performance. Experiments have also been conducted to verify the efficiency of the proposed method.

Keywords: JPEG-LS, Lossless compression, Golomb code, Golomb-Rice, Location map

1. Introduction

The JPEG-LS algorithm [1] is a lossless or near-lossless compression standard for still images. Most of the recent compression algorithms accepted the *Modeling and Coding* [2] concept to achieve an efficient compression. JPEG-LS uses the two-sided geometric distribution (TSGD) model [3] of the prediction error that is based on the median edge detector (MED) [4], and encodes this model by Golomb-Rice (GR) coding [5]–[6]. As a lossless compression algorithm, it performs quite effectively and is superior to most of the other algorithms including JPEG2000 [7]–[10]. A group of authors tried to improve the performance of JPEG-LS by introducing novel prediction methods. Baligar et al. [7] proposed a linear prediction method that minimizes the squared error and exploits quadtree coding. Bedi et al. [8] proposed a prediction method that detects the diagonal edges in addition to the vertical and horizontal edges. While the compression performances of both studies [7]–[8] exceed that of JPEG-LS, the computational complexities of their algorithms are far greater [9]. Kademi et al. [10] paradoxically showed the superiority of JPEG-LS. Kim et al. [11] proposed hierarchical average and copy prediction (HACP) scheme and showed the upper bound of significant bit truncation (SBT) coding. Combinations of multiple predictions [12]–[13] are also used to improve the compression rates. Masmoudi et al. [14] proposed a block-based lossless image compression for which finite-mixture models and adaptive arithmetic coding are used. Zhao et al. [15] and Starosolski [16] showed that the context-based adaptive lossless image codec (CALIC) [17] provides high compression rates within reasonable time frames, while JPEG-LS is effective enough and very fast. As a result, the JPEG-LS algorithm is still considered excellent in terms of the compression rate and the compression time.

Mobasserri et al. [18] proposed a method for the embedding of data into the JPEG bitstream, whereby Huffman-code mapping is used, and the data embedding is performed according to a reversible mapping of the unused codewords; notably, only a fraction of the JPEG codewords are actually used. Later, Qian and Zhang [19] improved their method, whereby the most important contribution of their methods from [18] and [19] is a reversible data hiding capability for the improvement of the compression rate; this reversibility is used in this paper for the replacement of long codewords. Ding et al. [20] proposed a modified Golomb coding for an asymmetric TSGD.

Even though JPEG-LS is almost perfect as a lossless algorithm, its performance can still be improved. The goal of this paper is the improvement of the JPEG-LS performance through a slight modification of the long codewords. JPEG-LS accepted the GR code as a main encoding algorithm, but this GR code can sometimes produce excessively long codewords. A number of techniques can be used to remedy the long codeword artifact of the GR code; that is, the JPEG-LS algorithm modifies these long GR codewords into a fixed-length code (the authors call this modification JPEG-LS GR), while an additional improvement is proposed in this paper. The GR artifact can also be resolved by utilizing the location map. The location map is a tool for the book-keeping of the side information that assures the reversibility of data hiding schemes [21]–[23]. It is mostly used to avoid underflow, overflow or decoding errors at the decoding stage.

This paper shows that the location map can be used for lossless data compression; moreover, it shows that long codewords can be replaced with shorter codewords. The contribution of this paper consists of three methods that reduce the size of the symbol length in

bits, as follows; replacing the fixed-length prefixes with location information, reducing the location map itself using a variable-length information, and reducing the suffix.

This paper is organized as follows. Section 2 briefly summarizes the concepts of GR coding and JPEG-LS GR coding. Section 3 revisits the concept of the location map, and includes an explanation of the application of location information for JPEG-LS GR coding; a simple method for the reduction of the location map size is also presented; along with a demonstration of improvements of the JPEG-LS GR method for specific cases. Section 4 presents an analysis of the experiment results, whereby the performances of the JPEG-LS GR method and the proposed method are compared. Section 5 concludes the paper.

2. Golomb-Rice coding vs. JPEG-LS coding

When an integer N is divided by another integer m , a quotient q and a remainder r are derived so that the number N can be represented as $N = q \cdot m + r$. As long as m is given, the GR code in the JPEG-LS is represented by [unary quotient q] + [one bit for a separator “1”] + [binary remainder r]. JPEG-LS does not encode the pixel value itself, but encodes the *prediction error* through the GR coding. The integer notation is defined here as follows: N_D is the digit expression (either in 8 bits or 16 bits), N_B is the binary expression of N_D , and N_U is the unary expression. For example, when N_D is $3_{(10)}$ for an 8-bit depth image, N_B is $00000011_{(2)}$ and N_U is $000_{(1)}$, and they are the same representations. Note that N_U has three zeros since N_D is 3. Similarly, when a given prediction error in a decimal number is $N_D = 73_{(10)}$, its binary representation in 8-bit format is expressed as $N_B = 01001001_{(2)}$. Let a divisor be $m = 2^k$, for example, where $k = 2$, q is the first $8 - k$ bits of N_B , which is $q = 010010_{(2)} = 18_{(10)} = 00000000000000000000_{(1)}$; and r is the last k bits of N_B , which is $r = 01_{(2)}$. In this case, the corresponding GR code is represented by the following format:

$$N_{GR} = 00000000000000000000|1|01_{(2)}$$

The output N_{GR} is of a pseudo-binary form. Namely, it looks like a binary sequence, but it is actually a combination of binary symbols. The GR code starts with 18 leading zeros that are equivalent to the unary representation of the quotient. The first “1” that separates the quotient and the remainder with two red bars is a separator. The last part of “01” is just the remainder r in binary format.

Another given number $73_{(10)}$ can be represented in a different form depending on the k value. Let k be 1 so that $q = 0100100_{(2)} = 36_{(10)} = 00000000000000000000000000000000_{(1)}$ and $r = 1_{(2)}$. As a result, its GR code is obtained according to the following format:

$$N_{GR} = 00000000000000000000000000000000|1|1_{(2)}$$

In this case, the total length of the GR code is 38 bits, which seems excessively long; therefore, this lengthy GR code N_{GR} is not used whenever $q \geq 23_{(10)}$, but a shorter code N_{LS} is used in a different format as follows:

$$N_{LS} = 000000000000000000000000|1|01001000_{(2)}$$

where the eight LSB bits are represented as $N_D - 1 = 72_{(10)} = 01001000_{(2)}$. Now, 32 bits (i.e., [23 bits of leading zeros] + [1-bit separator “1”] + [8-bits of $N_D - 1$]) are needed for N_{LS} rather

than 38 bits for N_{GR} . This representation mostly shortens the length of the GR code.

To achieve consistency with the GR code, 23 zeros are leading both the separator and the eight bits of $N_D - 1$ itself for the pixels quantized with 8-bit depth; then, maximally 32 bits are needed in any case. A gain by the JPEG-LS GR code is obtained when the quotient is large enough, and the gain of the JPEG-LS GR code is six bits, which is six bits less than the GR code in this example.

Such a gain, however, does not always occur with the use of JPEG-LS GR coding. For example, a given number $47_{(10)}$ can be represented as follows:

$$N_B = 00101111_{(2)}$$

For $k = 1$, its GR code is represented with 23 leading zeros, as follows:

$$N_{GR} = 000000000000000000000000|1|1_{(2)}$$

However, the JPEG-LS GR code should be used whenever $q \geq 23$. The JPEG-LS GR code is, therefore, represented as follows:

$$N_{LS} = 000000000000000000000000|1|00101110_{(2)}$$

It is obvious that the JPEG-LS GR code needs 32 bits while the GR code needs only 25 bits. Even though the JPEG-LS GR code is not always superior to the GR code, the former code is used in the JPEG-LS algorithm. The length of the JPEG-LS GR code that includes a prefix, a separator, and a suffix is always 32 bits. A combination of the [*prefix* and *separator*] of the JPEG-LS GR code is called the *replaceable JPEG-LS GR prefix* or *replaceable prefix* for short in this paper. This replaceable prefix consists of 24 bits that comprise 23 leading zeros plus the separator “1”. The 8-bit suffix is called *intact suffix* in this paper. If the prefixes are replaceable, their associated suffixes are intact suffixes.

Notably, most images are quantized with eight bits per pixel and eight bits per color (such as red, green, and blue); however, some accurate images are quantized with 16 bits per pixel and 16 bits per color. In this case, the JPEG-LS GR code consists of a 47-bit prefix, a 1-bit separator, and a 16-bit suffix; therefore, the replaceable JPEG-LS GR prefix is 48 bits for the 16-bit-depth images.

3. Proposed Method

Both the *run* and *regular* modes are used for JPEG-LS. The *run* mode encodes the run length, which is the count of the same continuous values. Either GR or JPEG-LS GR code is used for each codeword when the *regular* mode is utilized, and this *regular* mode is the sole focus of this paper.

In this section, three contributions are introduced. First, the replacement of the replaceable JPEG-LS GR prefix with location information can reduce the codeword size. Second, the size of the location information itself can be reduced. Third, the intact suffix code size can also be reduced.

3.1 Use of Location Information & Occurrence Gain

Location information is a kind of side information to guarantee reversibility. The encoder and decoder can perform the same job if they share the same location information. If the encoder modifies the pixel values, the modification can cause an overflow or underflow error. Obviously, in that case, the encoder skips the pixel to avoid the causing of an error, and the pixel should be marked in the location map. The decoder also skips the decoding when it encounters the marked position. A set of the marked position information is called the *location map*.

The first contribution of this paper is the replacement of the replaceable prefixes. This paper identifies the possibility that the JPEG-LS GR codeword can still be shortened using a simple method; that is, the location information is used instead of the replaceable JPEG-LS GR prefix. Even though the replaceable prefix is removed, the intact suffix is not altered. The position information of the replaceable prefix will be used here instead of the lengthy JPEG-LS GR code.

The location map consists of the simple (*row, col*) coordinates of such JPEG-LS GR code in this paper. The map of each *row* and *column* are of an 8-bit length in the 256 × 256 images, while a 9-bit length applies for the 512×512 images. For the 256×256 images, the coding gains for each replaceable prefix are therefore 8 (i.e., 24 – 2 × 8) bits for the 8-bit-pixel depth, and 32 (i.e., 48 – 2 × 8) bits for the 16-bit depth. For the 512×512 images, the coding gains for each prefix are 6 (i.e., 24 – 2 × 9) bits for the 8-bit depth and 30 bits for the 16-bit depth. This location information is subsequent to the total number information of these locations at the beginning of the bitstream.

Table 1. Example of the JPEG-LS codewords for a 512 × 512 image

Position	N_D	k	Prefix	Separator	Suffix
(0, 0)	33	2	00000000	1	01
***	***	***	***	***	***
(197, 256)	84	1	<u>000000000000000000000000</u>	<u>1</u>	01010011
***	***	***	***	***	***
(339, 211)	111	2	<u>000000000000000000000000</u>	<u>1</u>	01101110
***	***	***	***	***	***
(480, 290)	239	3	<u>000000000000000000000000</u>	<u>1</u>	11101110
***	***	***	***	***	***
(480, 394)	139	0	<u>000000000000000000000000</u>	<u>1</u>	10001010
***	***	***	***	***	***
(511, 511)	***	***	***	***	***

At a position (*row, col*) = (339, 211) of the 512×512 image in **Table 1**, for example, the JPEG-LS GR code is 000000000000000000000000101101110₍₂₎, and a replaceable prefix 0000000000000000000000001₍₂₎ is encountered. This prefix is removed during the proposed encoding and only the intact suffix of 01101110₍₂₎ (i.e., $N_D - 1$) remains. After removing the replaceable prefix, its position information of 101010011011010011₍₂₎, which is equivalent to (339, 211), is recorded at the header of the bitstream and a coding gain of six bits is obtained. The replaceable prefixes are underlined in **Table 1**.

Let H and W be the height and width of the image, respectively. As long as the value $\lceil \log_2 H \rceil + \lceil \log_2 W \rceil$ is smaller than 24 for the 8-bit depth, or less than 48 for the 16-bit depth, the replacement with the location map results in a coding gain; such a gain is called *occurrence*

gain in this paper. Of course, the number of replaceable prefixes should be delivered to the decoder as a part of the side information.

3.2 Map Gain

The second contribution is a possible reduction of the location information itself. The position information of the first replaceable prefix is marked as it is because the reference information regarding the position is not available; however, for the rest of the replaceable prefixes, the information regarding the previous position is available. The available information can be exploited for the attainment of further gain. In the example in **Table 1**, four replaceable prefixes appear where the positions are (197, 256), (339, 211), (480, 290) and (480, 394); here, the row positions {197, 339, 480, and 480} are in an ascending order. For the first position (197, 256) for a 512×512 image, 18 (i.e., 9 plus 9) bits are needed to record $011000101100000000_{(2)}$. The next position (339, 211) is $101010011011010011_{(2)}$; here, because of the fact that $339 \geq 256$, it is obvious that 339 is located in the second half (bottom part) of the image, and the subsequent row positions that are larger than 256 indicate that they are located in the second half. As a result, these rows are all marked as $1xxxxxxx_{(2)}$. Because both the encoder and the decoder recognize this fact, it is not necessary to encode the first bit of “1”, and only the 8-bit $xxxxxxx_{(2)}$ is needed without the “1” in this case; therefore, only the 17 (i.e., 8 plus 9) bits of $\underline{11100000}100100010_{(2)}$ are recorded as the location information of (480, 290).

If a replaceable prefix is in the second half of the image, one bit of gain can be attained in the recording of the next piece of location information; similarly, if an i -th replaceable prefix is placed in the fourth quarter, two bits of gain can be attained in the recording of the position of the $(i+1)$ -th prefix.

The column information in the same row can also be reduced by using a similar method because the columns of the same row are of an ascending order; therefore, the location information of (480, 394) in **Table 1** is $11100000\underline{10001010}_{(2)}$ in 16 (i.e., 8 plus 8) bits according to a referencing of the position information of (480, 290).

This type of gain is called *map gain* in this paper. Of course, the replaceable prefix positions occur randomly; therefore, such a gain is not always achieved.

3.3 K3 (or K10) Gain

The third contribution is a size reduction of the intact suffix. When the *replaceable prefix* is encountered for a pixel that is quantized by eight bits, the manifestation of the leading 23 zeros means that this sample is encoded by the JPEG-LS GR method, whereby the value of the quotient $q \geq 23$ and the suffix value is $N_D - 1$; that is, the replaceable prefix does not occur whenever $q < 23$. By utilizing this fact, the divisor information k can be exploited to reduce the size of the intact suffix. With respect to $k = 3$ regarding the 8-bit pixel values, for example, the value range of the first five bits of any q is from 0 to 31; as a result, N_B ranges from $00000xxx_{(2)}$ to $11111xxx_{(2)}$, where $xxx_{(2)}$ stands for three bits of “don’t care” binary numbers (i.e., remainders).

Regarding the quotient part, numbers from 0 to 22 do not appear when $q \geq 23$, and, in this case, numbers from 23 to 31 appear when k is 3; that is, only nine numbers are actually used. Thus, quotient values from 23 to 31 can therefore be reassigned to other values from 0 to 8, respectively. **Table 2** shows the reassigned number representation.

A modified quotient of 0 is equivalent to the unmodified quotient value of 23, a modified quotient of 1 is equivalent to the unmodified quotient value 24, and so on. Only four bits are consequently needed to represent the nine numbers from 0 to 8 ($0000_{(2)}$ to $1000_{(2)}$), rather than

the use of five bits to represent numbers from 23 to 31 ($10111_{(2)}$ to $11111_{(2)}$); moreover, one more bit can be saved here. Three bits are assigned to represent the numbers from 0 to 6, and four bits are assigned to represent 7 and 8.

Table 2. q representation method with $K3$ gain in 8-bit image

Unmodified q	Bit expression	Modified q	Recorded bits
23	10111	0	000
24	11000	1	001
25	11001	2	010
26	11010	3	011
27	11011	4	100
28	11100	5	101
29	11101	6	110
30	11110	7	1110
31	11111	8	1111

N_B is used rather than $N_B - 1$ to record the suffix in this case. The intact suffix of position (480, 290), for example, is $11101110_{(2)}$ in **Table 1**. The original $N_B = 11101110_{(2)} + 1 = 11101111_{(2)}$ can be split into two parts. If k is 3, then the unmodified-quotient part is $11101_{(2)}$ and the remainder part is $111_{(2)}$. The unmodified-quotient part of $11101_{(2)}$ (i.e., $29_{(10)}$) is modified to $110_{(2)}$ (i.e., $6_{(10)}$) by the subtraction of 23. Through this modification, the intact suffix is changed to $110111_{(2)}$, which requires six bits rather than eight bits, and saves two bits. This type of gain is called $K3$ gain in this paper.

The same number of bits can be saved for the values that are quantized by a 16-bit depth with $k = 10$. Then, the N_B ranges from $000000xxxxxxxx_{(2)}$ to $111111xxxxxxxx_{(2)}$ in normal case. Similarly, only 17 numbers from $101111xxxxxxxx_{(2)}$ to $111111xxxxxxxx_{(2)}$ are actually replaceable since the replaceable prefix does not occur whenever $q < 47$. As a result, the assignment of six bits is not required, but five or four bits are enough. Two bits can be saved by assigning short binary numbers from $0000_{(2)}$ to $1110_{(2)}$ (for the unmodified $47_{(10)}$ to $61_{(10)}$, respectively) or one bit can be saved by assigning $11110_{(2)}$ and $11111_{(2)}$ (for the unmodified $62_{(10)}$ and $63_{(10)}$, respectively) (See **Table 3**). This gain is called $K10$ gain in this paper.

Table 3. q representation method with $K10$ gain in 16-bit image

Unmodified q	Bit expression	Modified q	Recorded bits
47	101111	0	0000
48	110000	1	0001
49	110001	2	0010
...
59	111011	12	1100
60	111100	13	1101
61	111101	14	1110
62	111110	15	11110
63	<u>111111</u>	16	11111

4. Experiment Results

The performance of the proposed method was compared with that of the JPEG-LS algorithm (see **Tables 4 to 8**). A variety of uncompressed images ([24]-[25], **Figs. 1 and 2**) are used for the comparison; here, a few images do not have a replaceable prefix, while most of the images have a sufficient number of replaceable prefixes.

In the experiments, 12 gray-scale images of a 256×256 size and 12 gray-scale 512×512 images, all quantized by eight bits, are used; in addition, 8 color images of varying sizes that have been quantized by eight bits are used (see **Fig. 1**). The images quantized by 16 bits are shown in **Fig. 2**, and include 15 gray-scale images and 14 color images.

Before the compressed bitstream is recorded, the overhead representing the number of location information is first recorded.

For the *lena* (256×256) image with an 8-bit pixel depth, the replaceable prefixes occur 20 times, and 160 bits (i.e., 20×8) can be reduced purely by the *occurrence gain*. For the recording of the 20 occurrences, *five bits* are needed to represent the number $20_{(10)}$ or $10100_{(2)}$, and this number of bits, *five*, is expressed as $0101_{(2)}$. The entirety of the overhead information is therefore expressed with nine bits as $0101|10100_{(2)}$ and this overhead header is preserved at the beginning of the binary stream for this image.

Among the nine bits in the overhead header, the first four bits are used to represent the following number of bits representing the number of replaceable prefixes; therefore, the leading four bits can represent numbers from 0 ($0000_{(2)}$) to 15 ($1111_{(2)}$). The maximum number of overhead bits is 19 (four bits of leading “1111” plus 15 bits to represent the number of replaceable prefixes).

The exact output bitstream size of JPEG-LS depends on two kinds of zero padding. The first type pads a number of zeros at the end to make the bitstream size a multiple of eight; for example, when the actual output bitstream is $00000000110_{(2)}$ (11 bits), five zeros are padded to make 16 bits in total and the output is then $0000000011000000_{(2)}$. The second type of zero padding occurs whenever the consecutive eight bits are all “1”. For the bit stream $111111110_{(2)}$, where the first eight bits are all “1”, one zero is inserted as a delimiter and four zeros are padded to make $1111111011000000_{(2)}$. Such a zero-padding rule is briefly stated to calculate the exact number of the output bits of the JPEG-LS algorithm.

For the *lena* image (256×256), the original GR and JPEG-LS GR output stream size without any padding is 299,901 bits, and 123 bits are added through the padding process to make 300,024 bits; notably, since 300,024 is a multiple of eight, the first type of zero padding is not necessary. In the pure output bit string of 299,901 bits, 123 cases of $11111111_{(2)}$ occurs, and the same number of zero bits are inserted as a second type of zero padding.

In the proposed method of this paper, nine bits of overhead header (i.e., “ $0101|10100_{(2)}$ ”) are added to the pure JPEG-LS output of 299,901 bits; alternatively, 160 bits are reduced by the occurrence gain due to the 20 replaceable prefixes. In addition, six bits are reduced by the map gain; moreover, two bits are reduced by the K3 gain. The modified JPEG-LS output size is therefore 299,742 bits (i.e., $299,901 + 9 - 160 - 6 - 2$). Regarding the 299,742 bits, since there are 122 consecutive eight “1”s, 122 zeros are inserted. As a result, the final output size of the modified JPEG-LS is 299,864 (i.e., $299,742 + 122$) bits, which is the final result of the *lena* image because it is a multiple of eight, and this is slightly more favorable than that of the original JPEG-LS. In **Table 4, 5 and 7**, the numbers in the upper rows of the *occurrence gain* and the *K3 gain* (or *K10 gain*) are shown in bits, while the numbers in parenthesis in the lower rows indicate the number of occurrences of such gain.



Fig. 1. 8-bit small size images. (<http://links.uwaterloo.ca/Repository.html>)

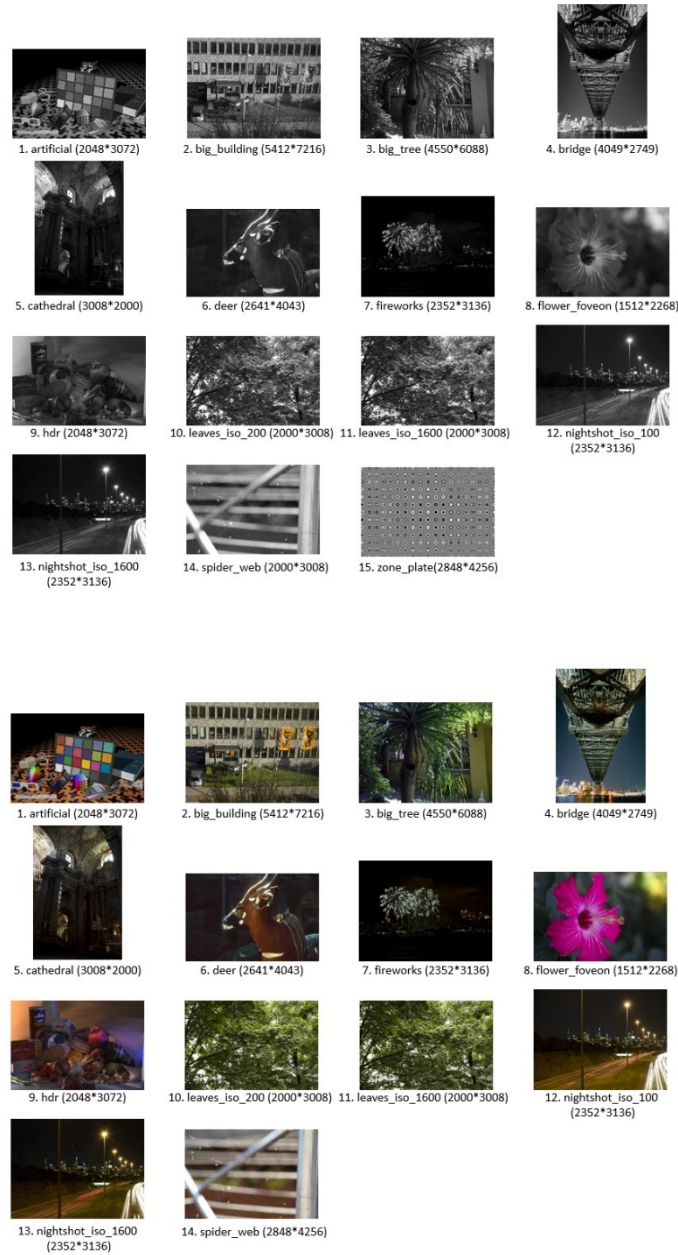


Fig. 2. 16-bit large size images. (http://imagecompression.info/test_images)

For a few images, replaceable prefixes are not encountered; however, some images have a large number of replaceable prefixes. The *artificial* image ($2,048 \times 3,072$) that has been quantized by 16 bits (see [Table 7](#)) has 452,775 bits of occurrence gain meaning that the artificial image has 18,111 replaceable prefixes. Since the image size is $2,048 \times 3,072$, 23 bits are needed to mark the positions; therefore, the apparent occurrence gain is 25 bits (i.e., $48 - 23$, where the replaceable prefix takes 48 bits for the 16-bit depth images). In this case, 19 bits are required as an overhead header to represent the 18,111 replaceable prefixes (i.e., the mandatory four bits to represent 15, and the next 15 bits to represent 18,111); therefore, the

overhead header is represented as 1111|10001101011111₍₂₎, which needs 19 bits in total. Notably, the *artificial* image is a synthetic image and not a natural one.

For the *cameraman* (256×256) image, 107 replaceable prefixes are encountered, and this means that 856 bits can be reduced; moreover, 32 bits are reduced due to the map gain, and 12 bits are reduced due to the K3 gain. As a result, the output size of the JPEG-LS and the proposed method are 282,488 bits and 281,576 bits, respectively. For the *France* image (496×672), 289 replaceable prefixes are found, while 112 bits and 33 bits are reduced as the map and K3 gains, respectively.

The *clegg* color image (880×814) in **Table 6** has 7,872 bits of occurrence gain for the red channel, 6,832 bits for the green channel, and 8,456 bits for the blue channel, since the image has 1,968, 1,708, and 2,114 replaceable prefixes, respectively. For the red color, 7,872 bits, 1,669 bits, and 536 bits are the occurrence gain, map gain, and K3 gain, respectively. For the images quantized with 16 bits, the K10 gain is a counterpart of the K3 gain, as shown in **Tables 7** and **8**.

Table 4. Result of 8-bit Gray Images I.

Image (H×W)	occurrence gain	map gain	K3 gain	output bits		
				Original	JPEG-LS	Proposed
bird (256×256)	8 bits (1)	0 bits	0 bits	524,288	227,272	227,264
bridge (256×256)	152 bits (19)	0 bits	0 bits	524,288	379,264	379,128
cameraman (256×256)	856 bits (107)	32 bits	22 bits (12)	524,288	282,488	281,576
circles (256×256)	24 bits (3)	0 bits	0 bits	524,288	9,784	9,768
corsses (256×256)	48 bits (6)	2 bits	0 bits	524,288	25,048	25,008
goldhill (256×256)	64 bits (8)	2 bits	2 bits (1)	524,288	345,896	345,840
horiz (256×256)	56 bits (7)	4 bits	0 bits	524,288	5,928	5,880
lena (256×256)	160 bits (20)	6 bits	2 bits (1)	524,288	300,024	299,864
montage (256×256)	504 bits (63)	34 bits	6 bits (3)	524,288	178,240	177,712
slope (256×256)	648 bits (81)	49 bits	8 bits (4)	524,288	102,760	102,064
squares (256×256)	24 bits (3)	0 bits	0 bits	524,288	4,840	4,824
text (256×256)	184 bits (23)	11 bits	0 bits	524,288	106,728	106,536

Table 5. Result of 8-bit Gray Images II.

Image ($H \times W$)	occurrence gain	map gain	K3 gain	output bits		
				Original	JPEG-LS	Proposed
barb (512×512)	132 bits (22)	11 bits	0 bits	2,097,152	1,240,584	1,240,448
boat (512×512)	84 bits (14)	2 bits	2 bits (1)	2,097,152	1,113,832	1,113,752
France (496×672)	1,445 bits (289)	112 bits	65 bits (33)	2,666,496	470,664	469,048
frog (498×621)	60 bits (12)	3 bits	2 bits (1)	2,474,064	1,870,432	1,870,368
goldhill (512×512)	24 bits (4)	1 bits	0 bits	2,097,152	1,234,912	1,234,896
lena (512×512)	72 bits (12)	6 bits	0 bits	2,097,152	1,112,240	1,112,176
library (352×464)	624 bits (104)	50 bits	37 bits (20)	1,306,624	832,936	832,240
mandrill (512×512)	492 bits (82)	0 bits	6 bits (3)	2,097,152	1,582,216	1,581,728
mountain (480×640)	885 bits (177)	99 bits	68 bits (39)	2,457,600	1,972,616	1,971,592
peppers (512×512)	126 bits (21)	6 bits	2 bits (1)	2,097,152	1,176,472	1,176,344
washsat (512×512)	36 bits (6)	1 bits	0 bits	2,097,152	1,082,256	1,082,232
zelda (512×512)	6 bits (1)	0 bits	0 bits	2,097,152	1,049,760	1,049,752

Table 6. Result of 8-bit Color Images.

Image ($H \times W$)	C	occurrence gain	map gain	K3 gain	output bits		
					Original	JPEG-LS	Proposed
clegg (880×814)	R	7,872	1,669	965	5,730,560	1,703,848	1,693,352
	G	6,832	1,045	515	5,730,560	1,794,640	1,786,264
	B	8,456	1,479	988	5,730,560	1,783,608	1,772,808
frymire (1105×1118)	R	4,268	64	1,185	9,883,120	2,431,024	2,425,608
	G	4,224	172	993	9,883,120	2,609,960	2,604,648
	B	3,818	108	906	9,883,120	2,459,416	2,454,648
lena (512×512)	R	18	0	0	2,097,152	1,060,280	1,060,272
	G	102	6	0	2,097,152	1,207,568	1,207,472
	B	192	0	8	2,097,152	1,284,120	1,283,944
monarch (512×768)	R	530	61	6	3,145,728	1,469,288	1,468,712
	G	445	45	14	3,145,728	1,469,056	1,468,576
	B	415	38	12	3,145,728	1,510,704	1,510,240
peppers (512×512)	R	114	7	1	2,097,152	1,034,904	1,034,784
	G	108	16	2	2,097,152	1,019,296	1,019,176
	B	150	16	2	2,097,152	1,029,288	1,029,120
sail (512×768)	R	110	3	4	3,145,728	2,060,152	2,060,040
	G	100	0	0	3,145,728	2,042,296	2,042,208
	B	95	5	0	3,145,728	2,050,112	2,050,024

serrano (794×629)	R	2,972	247	235	3,995,408	826,480	823,072
	G	3,104	319	304	3,995,408	867,944	864,240
	B	2,204	254	269	3,995,408	665,408	662,680
tulips (512×768)	R	155	9	0	3,145,728	1,578,680	1,578,520
	G	125	8	0	3,145,728	1,656,520	1,656,400
	B	160	15	2	3,145,728	1,701,432	1,701,272

Table 7. Result of 16-bit Gray Images.

Image (H×W)	occurrence gain	map gain	K10 gain	output bits		
				Original	JPEG-LS	Proposed
artificial (2048×3072)	452,775 (18,111)	22,765	6 (3)	100,663,296	27,167,728	26,692,336
big_building (5412×7216)	5,060 (230)	0	0	624,847,872	451,339,088	451,334,028
big_tree (4550×6088)	198 (9)	18	0	443,206,400	324,584,184	324,583,986
bridge (4049×2749)	120 (5)	12	0	178,091,216	135,643,808	135,643,688
cathedral (3008×2000)	1,225 (49)	84	2 (1)	96,256,000	69,526,592	69,525,296
deer (2641×4043)	240 (10)	0	0	170,841,008	136,309,472	136,309,240
fireworks (2352×3136)	4,464 (186)	101	0	118,013,952	60,353,296	60,348,744
flower_foveon (1512×2268)	1,375 (55)	20	0	54,867,456	33,892,768	33,891,376
hdr (2048×3072)	550 (22)	28	0	100,663,296	62,900,288	62,899,720
leaves_iso_200 (2000×3008)	3,350 (134)	125	4 (2)	96,256,000	70,434,464	70,431,000
leaves_iso_1600 (2000×3008)	3,075 (123)	168	4 (2)	96,256,000	74,955,104	74,951,864
nightshot_iso_100 (2352×3136)	648 (27)	9	0	118,013,952	72,684,152	72,683,504
nightshot_iso_1600 (2352×3136)	2,808 (117)	60	0	118,013,952	88,672,408	88,669,552
spider_web (2848×4256)	575 (25)	12	0	193,937,408	111,630,984	111,630,400
zone_plate (2000×3000)	1,675 (67)	30	56 (30)	96,000,000	95,673,640	95,671,872

Table 8. Result of 16-bit Color Images.

Image ($H \times W$)	C	occurrence gain	map gain	K10 gain	output bits		
					Original	JPEG-LS	Proposed
artificial (2048×3072)	R	440,400	22,646	14	100,663,296	27,622,520	27,159,592
	G	455,325	22,522	5	100,663,296	26,097,064	25,619,352
	B	392,650	23,018	6	100,663,296	22,666,056	22,250,512
big_building (5412×7216)	R	1,210	0	0	624,847,872	471,972,608	471,971,398
	G	7,568	2	0	624,847,872	446,166,904	446,159,336
	B	990	6	2	624,847,872	457,438,424	457,437,434
big_tree (4550×6088)	R	198	0	2	443,206,400	338,047,656	338,047,458
	G	1,254	6	0	443,206,400	316,958,648	316,957,394
	B	176	0	7	443,206,400	356,661,248	356,661,072
bridge (4049×2749)	R	1,056	37	0	178,091,216	137,191,176	137,190,088
	G	48	3	0	178,091,216	139,260,144	139,260,096
	B	0	0	0	178,091,216	141,927,112	141,927,120
cathedral (3008×2000)	R	2,125	172	6	96,256,000	72,049,384	72,047,096
	G	2,450	204	0	96,256,000	65,756,096	65,753,448
	B	3,025	352	0	96,256,000	69,606,856	69,603,488
deer (2641×4043)	R	72	0	0	170,841,008	136,085,256	136,085,192
	G	0	0	0	170,841,008	143,742,960	143,742,968
	B	0	0	0	170,841,008	147,174,488	147,174,496
fireworks (2352×3136)	R	3,864	63	0	118,013,952	65,342,088	65,338,176
	G	2,424	0	4	118,013,952	50,603,008	50,600,600
	B	7,104	27	0	118,013,952	30,611,824	30,604,712
flower_foveon (1512×2268)	R	50	0	0	54,867,456	34,634,848	34,634,800
	G	1,025	18	0	54,867,456	25,886,872	25,885,840
	B	0	0	0	54,867,456	34,599,072	34,599,072
hdr (2048×3072)	R	300	22	0	100,663,296	64,317,448	64,317,136
	G	250	15	0	100,663,296	64,468,512	64,468,256
	B	175	11	2	100,663,296	66,914,480	66,914,288
leaves_iso_200 (2000×3008)	R	2,550	118	15	96,256,000	71,013,008	71,010,336
	G	5,150	251	2	96,256,000	70,067,784	70,062,400
	B	1,700	58	8	96,256,000	69,239,256	69,237,488
leaves_iso_1600 (2000×3008)	R	425	11	9	96,256,000	76,392,784	76,392,352
	G	4,025	235	6	96,256,000	74,091,952	74,087,696
	B	450	13	2	96,256,000	73,353,248	73,352,800
nightshot_iso_100 (2352×3136)	R	1,080	9	0	118,013,952	74,671,984	74,670,904
	G	1,056	18	0	118,013,952	71,826,064	71,825,000
	B	1,944	48	0	118,013,952	60,798,592	60,796,608
nightshot_iso_1600 (2352×3136)	R	72	0	0	118,013,952	90,512,384	90,512,320
	G	2,928	9	0	118,013,952	86,662,496	86,659,568
	B	2,712	3	0	118,013,952	80,695,226	80,668,410
spider_web (2848×4256)	R	115	0	0	193,937,408	114,490,864	114,490,752
	G	529	8	0	193,937,408	111,667,024	111,666,496
	B	207	2	0	193,937,408	114,798,856	114,798,656

5. Conclusion and Discussion

In this paper, a new lossless compression method is proposed. When the codeword length is longer than the position information, the codeword can be replaced with a shorter piece of position information. JPEG-LS reference software uses either 24-bit or 48-bit replaceable prefixes to avoid the long codewords of the GR code; however, it is obvious that they are still long, and that replaceable prefixes of less than 24 bits or 48 bits can be used to enhance the compression rate. Future work includes the employment of various lengths of the replaceable prefix to maximize the performance. In this paper, the potential of the location information is manifested resulting in additional gains.

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