



# Image Compression and Decompression using Coding Redundancy Method

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*Abstract: Image compression is a technology in the field of digital image processing. It is a process of reducing the size of image at the time of storing on the disk and transmitting the image from one place to other place. Now days whole world adopt the online system for carry out any official or academic work for sending or receiving textual or image data. Therefore for fast speed transmitting, image should have smaller size. Different techniques are available for reducing the size of image. In this paper I discussed on coding redundancy method for lossless type of images, in which it choose a less number of bits for frequently occurring data at the time of compression. This paper explained a compression technique and its model with algorithm for an image which will easy for the programmer for coding or writing program for compression.*

*Keywords- Lossless, coding redundancy method, compression.*

## 1. Introduction

Image compression plays a key role in much application such as image data base, image communication, digital movie, televideo conferencing, remote sensing ,document and medical imaging, fax and in camera. It speed the processing due to reduce the size of image. Web page image and high resolution digital camera photos also are compressed routinely to save storage space and reduce transmission time. The residential internet connections deliver data at speed ranging from 56 Kbps via conventional phone line and more than 12 Mbps for broadband. The time required to transmit a small 128X128X8 bit full color image over this range of speed is from 7.0 to 0.03 seconds. Compression can reduce the transmitting time by the factor of 2 to 10 or more. Number of works has been done on this area. In [4] Syed Ali Hassan and Mehdi Hussain described Lossless Image Data Compression Method in spatial domain. In [5] author has described a lossless image compression algorithm using duplication run length coding.

Saravanan et. al.[6] explained lossless grey-scale image compression using source symbol reduction and Huffman Coding. Effective context lossless image coding approach based on adaptive prediction is explained in [9].Xiwen Zhao has explained a local structure learning and prediction for efficient lossless image compression. Lossless image compression using super-spatial structure prediction is explained in 11.

This paper described the model of compression using Huffman code technique which is based on Coding redundancy method and also designed a small algorithm for compressing and decompression of an image.



## 2. Background

Data compression is the process of reducing the amount of data required to represent a given quantity of information. Data are the means by which information is conveyed. Amount of data can be used to represent the same amount of information, representation that contain irrelevant or repeated information are said to contain redundant data. Suppose  $b$  and  $b'$  contain the number of bits in two representation of the same information then compression ratio is nothing but ratio of  $b$  and  $b'$  and relative data redundancy ( $R$ )

$$R = 1-1/C$$

Which shows how much percent data is redundant. Where  $C$  is called compression ratio defined as

$$C = b/b'$$

### 2.1 Different types of redundancy

There are different types of redundancies that can be identified in two dimensional representation of image

- a) Coding redundancy- A code is a system of symbols used to represent a body of information or a set of events. Each piece of information or event is assigned a sequence of code symbol called a code word. The number of symbol in each code word is its length. In coding redundancy the symbol which are used many times in a image, set a less number of symbols. In this way reducing the size of image.
- b) Spatial redundancy-this types of redundancy called a inter pixel redundancy. The neighborhood pixels are correlated. So the redundant data may be deleted to compress image.
- c) Irrelevant information- Most 2D intensity arrays contains information that is ignored by the human visual system. That information is redundant data.

Data compression is achieved when one or more of these redundancies are eliminated. Hoffman coding method uses a coding redundancy to compress the image.

## 3. Types of compression

There are two types of compression. Lossless compression and lossy compression. In lossless compression compress image is identical to the original image while lossy compression loss some information of the image and little bit change in the decompressed image. Due to telemedicine facility now a days, medical image may be transmitted fast due to compressed and may be diagnosed anywhere in the country in short period of time. Lossless image compression is highly suitable for medical image. Compression techniques for the lossless compression are

- a) Run length encoding
- b) Huffman Coding
- c) Arithmetic coding

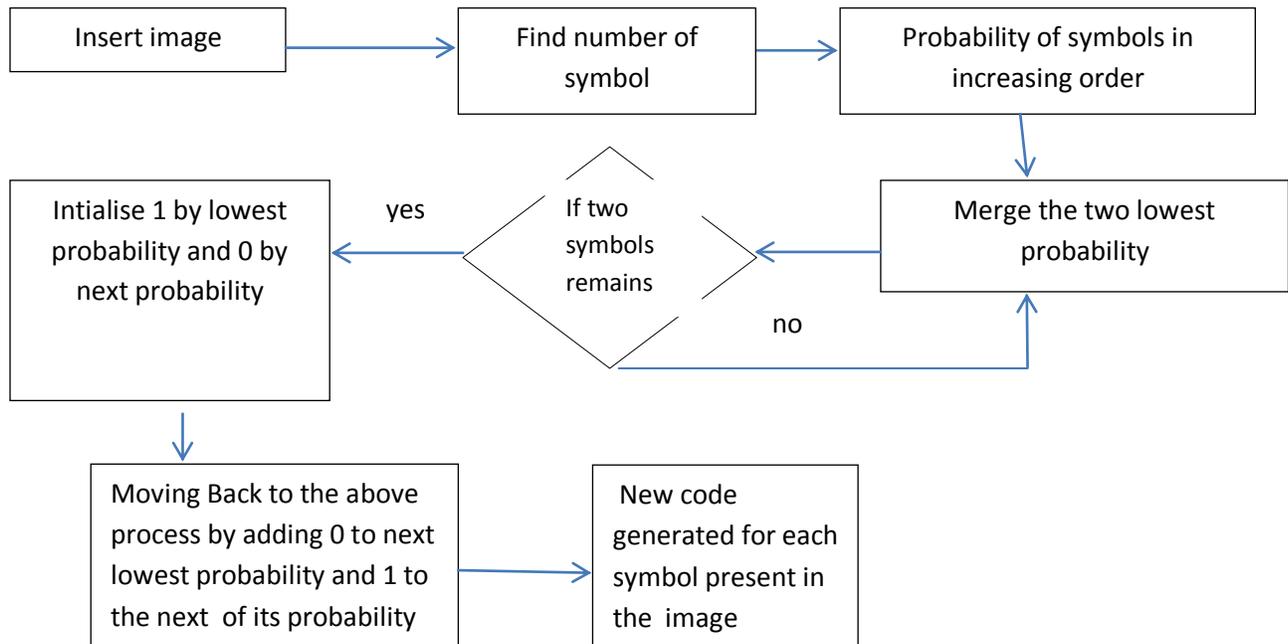
- d) Entropy coding
- e) Area coding

Lossy image loss some information so it is suitable only for the photographic image.  
Compression techniques for the Lossy compression

- a) Predictive coding
- b) Transform coding(FT/DCT/Wavelet)

#### 4. Coding redundancy technique

Huffman coding method of compression uses a coding redundancy technique. In this technique image is first digitized and sampled.. In the 8-bit representation each pixel use 8 bit to represent intensity in the image. Probability is used to calculate largest occurrence of intensity present in the image. This method reduces the number of bits for largely occurrence symbol. In this way storage capacity of image may reduce. Figure 1 shows the model of compression of an image



**Figure 1. Models of Huffman Coding as Encoder**



Let us take one example to understand the Huffman coding technique-

Probability of each symbol can be calculated by  $n_{a_k}/n$  where  $n_{a_k}$  is number of times  $a_k$  occurs in the image and  $n$  is the total number of pixel in image. If  $N$  is number of row and  $M$  is numbers of column then  $n= MN$ .

Entropy can be the average number of bits used to represent each pixel. Let each symbol or each pixel uses 8 bits then entropy can be calculated as

$$\text{Entropy or } L \text{ avg} = \sum l(a_k) \text{ pr}(a_k) \text{ for } k= 0 \text{ to } n-1$$

Where  $l(a_k)$  is number of bits used to represent the symbol  $a_k$ . Table 1 shows the variable length code of each symbol present in the image with probability is mentioned. Table 1 shows the process of finding the codes of each symbols presents in the image by source reduction technique.

**Table 1. Huffman Source Reduction**

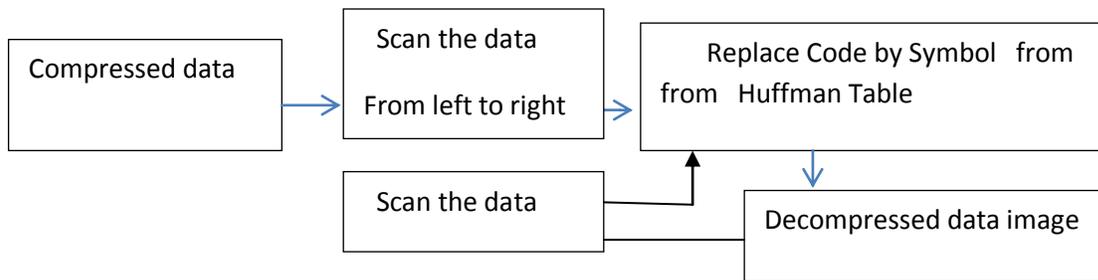
| symbol | probability | Code | Source reduction technique |            |           |           |     |   |
|--------|-------------|------|----------------------------|------------|-----------|-----------|-----|---|
| a1     | 0.5         | 01   | 0.5<br>01                  | 0.5<br>01  | 0.6<br>00 | 0.8<br>1  | 1.1 | 0 |
| a2     | 0.4         | 10   | 0.4 10                     | 0.4<br>10  | 0.5<br>01 | 0.6<br>00 | 0.8 | 1 |
| a3     | 0.3         | 000  | 0.3 000                    | 0.4<br>11  | 0.4<br>10 | 0.5<br>01 |     |   |
| a5     | 0.3         | 001  | 0.3 001                    | 0.3<br>000 | 0.4<br>11 |           |     |   |
| a8     | 0.2         | 110  | 0.2 110                    | 0.3<br>001 |           |           |     |   |
| a7     | 0.1         | 1110 | 0.2<br>111                 |            |           |           |     |   |
| a6     | 0.1         | 1111 |                            |            |           |           |     |   |

Probability of each symbol is depicted in ascending order. This method is static in nature so whole image is scanned first for further processing. After scanning complete image, probability of each symbols is initialized. Next step is to combine the lowest probability symbols in to compound symbols that replace them in to next source reduction. This compound symbol and associated probability are placed in the first source reduction column. This method is repeated until two symbols are reached. Next steps in this method are to provide variable length code for

each symbols starting with the smallest source and working back to the original source. In the 5<sup>th</sup> column of source reduction smallest number is 0.8 so code for this is taken as 1. And 1.1 is the combination of 0.6 and 0.5 so 0.6 is taken as 00 and 0.5 is taken as 01.next 0.8 is the combination of last two lowest probabilities 0.4 and 0.4 so coding is proved accordingly as 10 and 11. Repeat the same process until reached to the first source reduction. Here a1 has more probability but less number of bits. At the time of scanning image row by row, binary bits are replaced by particular symbolic binary digits mentioned in the table 1 and send or store in the internet or disk which reduce the transmission as well storage capacity. After complete compression of the image data are stored in the terms of bits in the file. .

### 5. Model for Compression and Decompression of an image

Input image  $f(x,y)$  is fed into encoder i.e. Huffman encoding procedure, which create a compressed representation of the input. This representation is stored for later use or transmitted at a remote location. Now compressed representation is presented on the decoder, At the time of decompression particular symbolic binary code is replaced by original code bits which is done by a decoder and original image is obtained at receiving end.



**Figure-2: Decompression of an Image using Coding Redundancy**

### 6. Conclusion and Future Cope

We presented a lossless image compression using code redundancy technique. We achieved 60% reduction in gray scale images so finding fast transmission time and less storage capacity. Mean square error has the value zero and Peak signal to noise ratio has the value infinite because it is the inversely proportional to mean square error. In future we will apply this method to colored image compression and video compression and medical image of CT scan type MRI and X-Ray image. Huffman technique is also used nowadays for compression of image dynamically, coming online data compress in parts one by one.



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