

## JPEG Encoder using Discrete Cosine Transform & Inverse Discrete Cosine Transform

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**Abstract:** In the past decade, the advancement in data communications was significant during explosive growth of the Internet, which led to the demand for using multimedia in portable devices. Video and Audio data streams require a huge amount of bandwidth to be transferred in an uncompressed form. The objective of this paper is to minimize the number of bits required to represent an image and also the accuracy of discrete interpolation by using DCT. Here we followed JPEG standard for compression by using DCT because of its energy compaction property. The original image is taken through a series of steps, which include initial input, DCT, Quantization and Encoding. For different quality factors, the compression ratios were also calculated.

**Keywords:** Compression Ratio, DCT, Huffman Coding, JPEG Encoder, Quality factor, Quantization, Zig Zag.

### I. Introduction

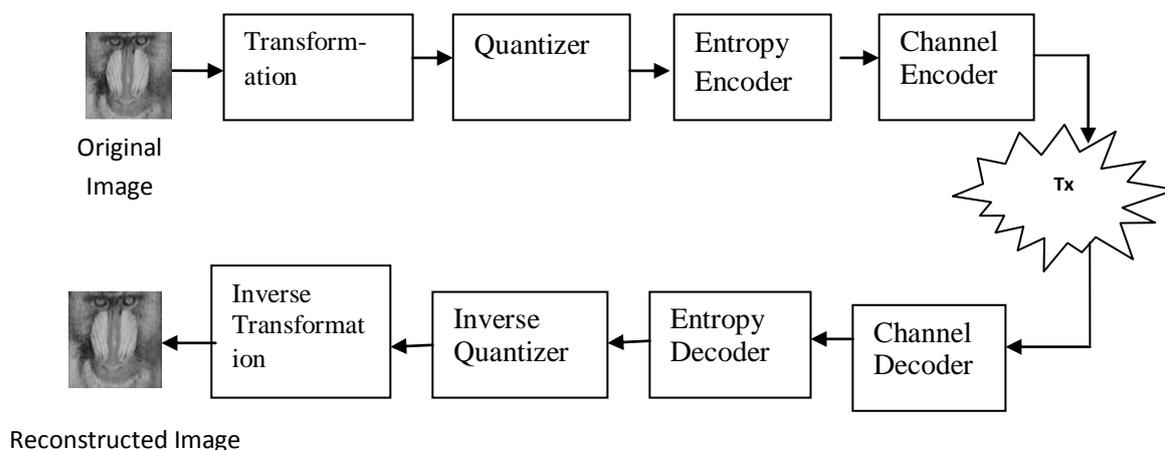
Image compression refers to the process of reducing the amount of data required to represent an image without any appreciable loss of information. If the amount of the data associated with image is large, then a large channel capacity is required. For motion and color pictures, storage capacity will increase momentarily.

Transmission and storage of uncompressed video/image would be extremely costly and impractical. A frame with 352x288 contains 202,752 bytes of information. Recoding of uncompressed version of this video at 15 frames per second would require 3 MB. One minute → 180 MB storage. One 24-hour day → 262 GB.

Using compression, 15 frames/second for 24 hour → 1.4 GB, 187 days of video could be stored using the same disk space that uncompressed video would use in one day.

DCT is the compression technique, which reduces the number of bits required to represent an image. The implementation of this technique considers the principle which states that the human eye is more sensitive to high frequencies than low frequencies. Hence we neglect high frequency components in the image using the non-uniform quantization technique. By applying different coding techniques we reduce the number of bits to represent an image. The block diagram of Image compression is shown in the below figure(1).

Fig 1. Block Diagram of Image Compression



### II. Jpeg Encoder

JPEG (pronounced "jay-peg") is a standardized image compression mechanism. JPEG stands for Joint Photographic Experts Group, the original name of the committee that wrote the standard. JPEG is designed for compressing either full-color or gray-scale images of natural, real-world scenes. It works well on photographs, naturalistic artwork, and similar material not so well on lettering, simple cartoons, or line drawings. JPEG handles only still images, but there is a related standard called MPEG for motion pictures. JPEG is "lossy,"

meaning that the decompressed image isn't quite the same as the one that started with. A useful property of JPEG is that the degree of lossiness can be varied by adjusting compression parameters.

Figure 2.1 describes the JPEG process. JPEG divides the image into 8x8 pixel blocks and then calculates the DCT of each block. A quantizer rounds off the DCT coefficients according to the quantization matrix. This step produces the “lossy” nature of JPEG, but allows for large compression ratios.

JPEG’s compression technique uses a variable length code on these coefficients, and then writes the compressed data stream to an output file (\*.jpg). For decompression JPEG recovers the quantized DCT coefficients from the compressed data stream, takes the inverse transforms and displays the image.

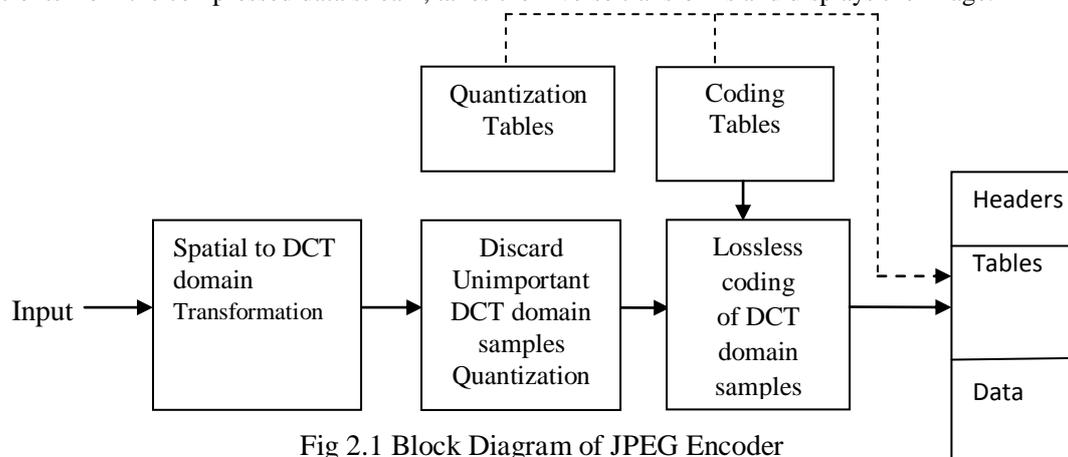


Fig 2.1 Block Diagram of JPEG Encoder

### III. D Dct/Idct Architecture And Jpeg

The key to the JPEG compression is a mathematical transformation known as the Discrete Cosine Transform (DCT). It also includes the well-known Fast Fourier Transform (FFT). Its basic operation is to take a signal and transform it from one type of representation to another.

In this case the signal is a graphical image. The concept of this transformation is to transform a set of points from the spatial domain into an identical representation in frequency domain. It identifies pieces of information that can be effectively —thrown away without seriously reducing the image’s quality.

The mathematical function for a two-dimensional DCT is:

$$DCT(i, j) = \frac{1}{\sqrt{2N}} C(i)C(j) \sum_{x=0}^{N-1} \sum_{y=0}^{N-1} pixel(x, y) \cos\left[\frac{(2x+1)i\pi}{2N}\right] \cos\left[\frac{(2y+1)j\pi}{2N}\right] \quad 3.1$$

$$C(x) = \frac{1}{\sqrt{2}} \text{ If } x \text{ is } 0, \text{ else } 1 \text{ if } x > 0 \quad 3.2$$

The DCT operation is shown in the below example:

The input 8 x 8 matrix (Fig 3.1) from a gray-scale image consists of pixel values which are randomly spread around 140 to 175 range. These values are fed to the DCT algorithm, creating the output matrix (Fig 3.2).

The output matrix consists of DCT Coefficients, which is ordered in a way that coefficients containing useful and important data for representation of the image are in the upper left of the matrix and in the lower right coefficients containing less useful information.

The DC coefficient (represented in bold as shown in fig 3.2) is at position (0, 0) in the upper left-hand corner of the matrix and it represents the average of the other 63 values in the matrix. This step prepares the image for the next step, namely quantization.

140	144	147	140	140	15	179	175
144	152	140	147	140	148	167	179
152	155	136	167	163	162	152	172
168	145	156	148	140	136	174	162
162	148	156	148	140	136	147	162
147	167	140	155	155	140	136	162
136	156	123	167	162	144	140	147
148	155	136	155	152	147	147	136

Fig 3.1 Input Matrix

<b>186</b>	-18	15	-9	23	-9	-14	19
21	-34	26	-9	-11	11	14	7
-10	-24	-2	6	-18	3	-20	-1
-8	-5	14	-15	-8	-3	-3	8
-3	10	8	1	-11	18	18	15
4	-2	-18	8	8	-4	1	-4
9	1	-3	4	-1	-7	-1	-2
0	-8	-2	2	1	4	-6	0

Fig 3.2 Output Matrix

For decoding purpose there is an Inverse DCT (IDCT):

$$pixol(x, y) = \frac{1}{\sqrt{2N}} \sum_{i=0}^{N-1} \sum_{j=0}^{N-1} C(i)C(j)DCT(i, j)Cos\left[\frac{(2x+1)i\pi}{2N}\right]Cos\left[\frac{(2y+1)j\pi}{2N}\right] \quad 3.4$$

$$C(x) = \frac{1}{\sqrt{2}} \text{ if } x \text{ is } 0, \text{ else } 1 \text{ if } > 0 \quad 3.5$$

#### IV. Quantization & Entropy Coding

##### 4.1 Quantization

Till now, in the JPEG compression, partial image compression has occurred. Of all the steps quantization is the essence of lossy compression.

Quantization is a process of reducing the number of bits needed to store an integer value by reducing the precision of the integer. For every element in the DCT matrix, a corresponding value in the quantization matrix gives a quantum value. The quantum value is that which is stored in the compressed image.

The formula for quantization is:

$$QuantizedValue(i, j) = \frac{DCT(i, j)}{Quantum(i, j)}, \quad (\text{Rounded to the nearest integer}) \quad 4.1$$

Selecting a quantization matrix must be done carefully. Although JPEG allows for the use of any quantization matrix, ISO has done extensive testing and developed a standard set of quantization values that cause impressive degrees of compression.

To calculate the quantization matrix we use this formula:

**Quantization matrix Q (i, j) is: Q (i, j) = 1 + (1+i+j). (Quality Factor) 4.2**

The next step in the JPEG process is coding the quantized images. This is done through following steps:

1. Convert the DC coefficient to a relative value.
2. Reorder the DCT block in a zig-zag sequence as shown in fig 4.1.

Because so many coefficients in the DCT image are truncated to zero values during the coefficient quantization stage, the zeros are handled differently than non-zero coefficients. They are coded using a Run-Length Encoding (RLE) algorithm.

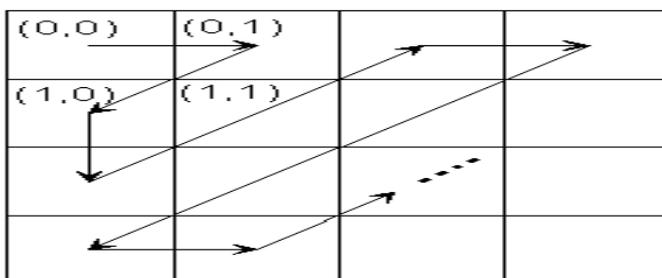


Fig 4.1: Zig-Zag sequence for Binary Encoding

##### 4.2 Entropy Coding

Finally, the JPEG algorithm outputs the DCT block's elements using an entropy encoding mechanism that combines the principles of RLE and Huffman encoding. The output of the entropy encoder consists of a sequence of three tokens, repeated until the block is complete.

The three tokens are:

- (1) The *run length*, the number of consecutive zeros that precede the current zero element in the DCT output matrix;
- (2) The *bit count*, the number of bits used to encode the amplitude value that follows, as determined by the Huffman encoding scheme; and
- (3) The *amplitude*, the amplitude of the DCT coefficient

##### 4.2.1 Coding the DC Coefficients

Firstly, we change the absolute value of the DC Coefficient at (0,0) to a relative value. Since adjacent blocks in an image display a high degree of correlation, coding the DC element as the difference from the previous DC element typically produces a very small number. This value is called the "DPCM difference" or the "DIFF" DC coefficients are sent with differential coding and is given by  $Z_i = DC_i - DC_{i-1}$

## 4.2.2 Coding the AC Coefficients

AC coefficients are coded in Zigzag (called in ZZ in standard) order to maximize possible runs of zeros. Code unit consist of run length followed by coefficient size. Baseline coding of size category is the same as for DC differences.

The following two algorithms are used for *Huffman coding*:

### I-Algorithm

#### 1. JPEG DC Coefficient Code Word Finding (DPCM)

(1). DC Coefficient = - 9., (2).Cat=4 [10], (3).Base Code =**101** and Length of code word=**7[10]**

(4) -9's, Binary Code=**0110** -----> it is called as LSB. (i.e., 9's Binary Code=1001 and it is 1's compliment=0110 and Add 1 ==>0111 ==> again sub 1=**0110**). i.e. Binary Code =>1's Comp+1=>sub 1==>LSB (if DC Coefficient -ve only other wise, we can take directly binary equivalent of DC Coefficient is LSB)

(5).Concatenate BC & LSB=CW from step 3 and 4

**DC Code Word= 101 0110**

### II-Algorithm

#### 2. JPEG AC Coefficient Code Word finding

(1). AC Coefficient = - 4, (2).Cat=3 [10], (3).run=0 from AC Coefficients Array; ==>run/Cat=0/3 ,(4).Base Code for 0/3 is =100 [10], (5) -4's, Binary Code=**011** -----> it is called as LSB (4's Binary Code=100 and its 1's compliment=011 and add 1=>**100** again sub 1=011).i.e. Binary Code =>1's Comp+1=>sub 1==>LSB (if AC Coefficient -ve only other wise, we can take directly binary equivalent of AC Coefficient is LSB ).

(6).Concatenate BC & LSB=CW from step 4 and 5

**AC Code Word= 100 011**

## V. Software Implementation & Compression Ratio

This application is developed in "MATLAB" on windows run PC. The input image file was taken as a BMP file of size 256x256 the output reconstructed image file also has the same size, but not the same quality, i.e. with a little visible distortion. The input image and reconstructed image files are shown in the next section. The Compression Ratio and Data Redundancy have been calculated for the compressed image data. As the compression ratio is increased, the quality of the image decreases. This is the limitation of JPEG.

The following details are the results of JPEG Compression: Input image file, output compressed image data, Reconstructed image, compression Ratio, Relative Data Redundancy are given below.

### 5.1 Input Image File Details

For implementation purpose, an image is considered, for which the source code has been written in Matlab. The input image is taj.bmp. All input files are in bit map format (BMP). Images can be taken in any format like bmp or JPEG.

#### 5.1.1 Details of Image Space

Name of image	:	a1.bmp
Color type	:	color (24-bit color image)
Compression	:	no compression
Resolution	:	96 pixels/inch
Size	:	256 X 256 X 24
Units	:	pixels
Input image data	:	$n_1=1572864$
Output compressed		
Image data	:	$n_2=373601$
Compression ratio	:	$C_r = n_1/n_2=1572864/373601=4.21$ i.e.
		approximatey 4:1

### 5.2 Relative Data Redundancy

Now calculating relative data redundancy

$Rd=1-1/cr$ , For image,  $Rd=1-1/4=.75=75\%$ .

Therefore, JPEG data now, 75% reduction i.e a compression ratio 4:1 means that the first data set has 4 information carrying units (say, bits) for every 1 unit in the second or compressed data set. The corresponding redundancy of 0.75 implies that the 75% of the data in the first data sent is redundant.

### VI. Result Analysis

In general, image compression means, by using less number of bits (compressing the bits) to reconstruct the original image with little visible distortion. Due to this distortion, various parameters of original image has changed Analysis of these parameters are shown in Table 1. The output result of reconstructed image files are shown below which have the final result. The output image is same as input image, but with little visible distortion as shown in the figures. The images of output files and input files have been enclosed. The data redundancies for different compression ratios are shown in Table 6.1

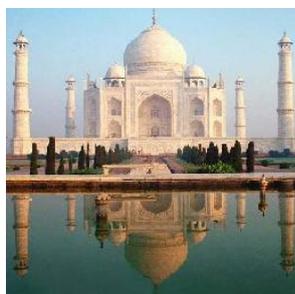


Fig 6.1 Original Image QF=9

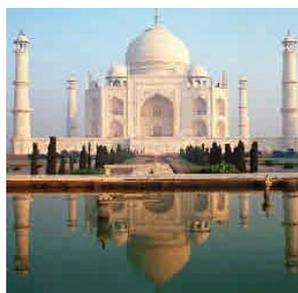


Fig 6.2 Compressed Image with QF=6.5



Fig 6.3 Compressed Image with QF=9.0

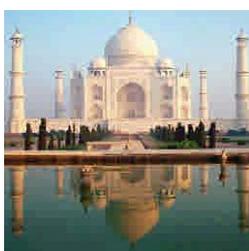


Fig 6.4 Compressed Image ,QF=14 QF=25



Fig 6.5 Compressed Image, QF=19

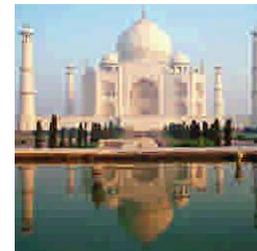


Fig 6.6 Compressed Image,

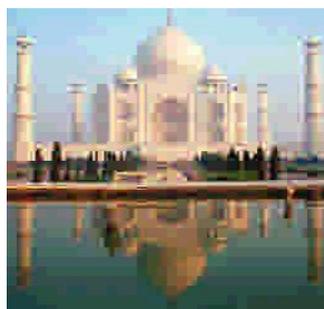


Fig 6.7 Compressed Image with QF=35



Fig 6.8 Compressed Image with QF=45

**TABLE 1. Comparison of different parameters of the output results**

S.No	Fig Name	Quality Factor	Actual Length	Compressed Length	Compression Ratio	Data Redundancy(%)
1	Fig 6.1	---	192 KB	---	---	-
2	Fig 6.2	6.5	192 KB	14.626 KB	13.13	92.3
3	Fig 6.3	9.0	192 KB	11.589 KB	16.57	93.9
4	Fig 6.4	14	192 KB	8.573 KB	22.40	95.5
5	Fig 6.5	19	192 KB	7.075 KB	27.51	96.3
6	Fig 6.6	25	192 KB	6.018 KB	31.91	96.8
7	Fig 6.7	35	192 KB	5.065 KB	37.90	97.3
8	Fig 6.8	45	192 KB	4.486 KB	42.80	97.6

## **VII. Conclusion And Future Scope**

Image compression is an extremely important part of modern computing. By having the ability to compress images to a fraction of their original size, valuable (and expensive) disk space can be saved. In addition, transportation of images from one computer to another becomes easier and less time consuming (which is why image compression has played such an important role in development of the internet). The JPEG image compression algorithm proves as a very effective way to compress images with minimal loss in quality. This work can be extended to compress the images by using the JPEG encoder for FPGA implementation by using suitable hardware. Using the Verilog HDL coding, the results may be compared with the above outputs which are implemented by using MATLAB. The accurate results may be noticed in terms of cost and power consumption.

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