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Assignment report Title: Technical Report

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Actual hrs spent for the assignment: 6 hours each

Assignment evaluation (3 key points):

- Getting some libraries to work in python was a pain as we had to reinstall python in an earlier version
 - Learning the jpeg algorithm was a blast for both of us
 - James' advice for the creation of the algorithm was very helpful

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Abstract

The purpose of image compression is to reduce the file-size of the image by getting rid of redundant data. Redundant data is defined by certain details in an image that if they were to be removed the human eye wouldn't be able to notice the difference. Applying image compression allows for files to occupy less storage on a device as well as require less bandwidth when sending data (in this context images) over a network.[1]

The purpose of this report is to highlight the effects of lossy and lossless compression applied to images.

Introduction

Image compression is one of the most important concepts in the modern era. Constantly, images and videos are being sent around the world. The problem that occurs is that all these media sources require a large amount of data to be transmitted from one device to another. Most devices are restricted with a limit to the amount of data they can receive and send at a given moment. A solution to this problem is compression. Compression is a term used to describe the act of pressing something into a smaller space [2]. In terms of computer systems, compression can reduce file sizes. Compression falls into two categories, lossy and lossless. Lossy compression is when the data or information is removed permanently and cannot be recovered, resulting in a reduction in file sizes and quality. It is most used for image and video compression. Lossless compression, on the other hand, is a form of compression that reduces the size of the file without removing the data permanently. With lossless, it is possible to reconstruct the original file perfectly after decompression. With the combination of these, images and video file sizes can be greatly reduced when being transmitted over a network. This is when JPEG compression is utilised and most applied.

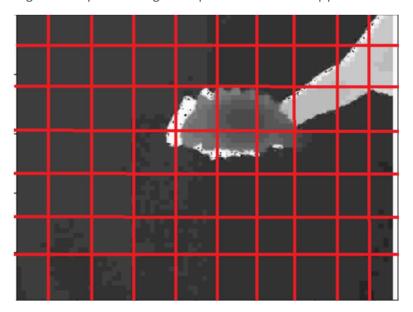
Within this report, we will be discussing and implementing the methods and algorithms of JPEG to compress and decompress a variety of different images. Metrics such as file size, MSE and compression ratio will be used as statistics to visualise the effects of the JPEG algorithm. This will be made possible using a user interface at which an image can be loaded, compressed and then decompressed into and out of memory.

Methodology

Theory

JPEG algorithm is used to compress images by getting rid of redundant data. The algorithm works by first dividing the RGB of the image data into luminance and colour components. The components are then divided into pixel blocks, with the block size normally being 8x8. The reason for that is because a block of size 4x4 would create poor quality (we define quality by calculating how different the decompressed images are from their original form) in the image when decompressed, while anything above a size 8x8 would create a higher quality image but at the cost of performance issues [3]. Figure 1 represents how the image is divided into pixel blocks

Figure 1 Representing how pixel blocks are applied to an image



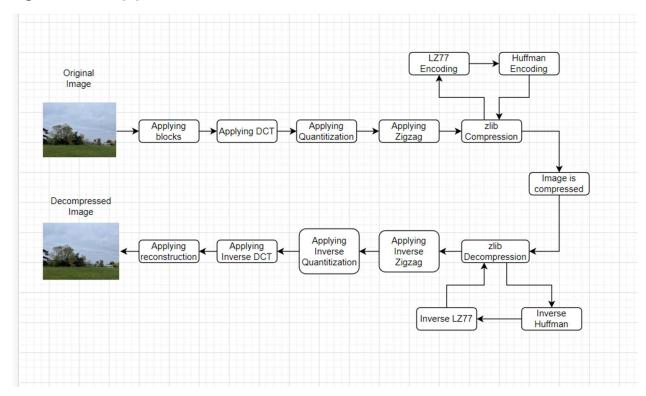
Each pixel block is then transformed using discrete cosine transformation (DCT). The image data in each block is represented using pixel numbers and values, after DCT the image data is represented as coefficients of the spatial domain frequencies for vertical and horizontal orientation. The next step is quantization. Quantization works by having each DCT element divided by the respective element inside a quantization table. Each application can supply its own values inside the quantization table as there is no standard, allowing control of the loss-compression trade-off depending on the needs of the application. This is done to get rid of redundant data or image data that wouldn't be noticed by the human eye if it was removed. The elements are then linearized using the zigzag method, and then run length and Huffman encoding are applied respectively. Both are types of lossless data compression algorithms.[4][5]

For decompression of the image, the explained method above is applied in reverse, producing an image with less file size but reduced quality. The reduced quality is dependent on different factors such as the size of the pixel blocks and the values inside the quantization table, however, under normal conditions, the reduction in quality shouldn't be easily recognizable by the human eye.

For this project, we have followed the format of the JPEG algorithm but instead of run-length encoding, we have used zlib, which is a software library written in C, used for data compression which utilizes the DEFLATE data compression format [6]. The DEFLATE format uses a combination of Huffman encoding and LZ77, which is another lossless data compression algorithm [7][8]. The use of the zlib software library has proven to be more efficient on our machines than implementing run-length and Huffman encoding separately.

Implementation details:

Figure 2. Shows the design of our compression and decompression algorithm, inspired by the JPEG algorithm from [4]



The design above has been implemented with PyCharm: python 3.9, using the following python libraries: *openCV; Scipy; Os; Numpy; Zlib; Matplotlib; warnings*

Functions

Function name	Purpose
encoding_dct_matrix()	This function applies the block padding of the image, applies DCT, quantization and zigzag then returns an encoded value.
compress()	Call all the functions needed to compress and decompress the image and display the image and all the statistics of the image.
CompressMenu()	Responsible for getting the user input regarding block size and quality factor and passing them to the function compress()
get_file_size()	Takes the images as a parameter and returns the file size in MB
menu()	Contains all the UI elements needed to allow the user to select, compress and display details about images.
decode_dct()	Does the inverse of all the methods that the <i>encoding_dct_matrix</i> does return the decompressed image
encoding_compressed_image()	This function applies lossless compression using the zlib compression software library.
decoding_compressed_image()	This function applies the inverse of the compression algorithms used by the zlib compression software library.
zigzag()	This applies the zigzag method after the image has been quantized.

Design of the Data Structures

Used numpy arrays to hold and manipulate the images . Used numpy integers for conversion of images. Dictionaries to hold all the images for user selection and 2D arrays to hold quantisation matrices. All the data structures are present in the code under the *Appendix*.

Results/Testing/Achievements

The program consists of a console-based user interface that allows the user to select an image, display the image, compress/decompress the image and switch between different images.

As we are saving and displaying the image as BMP, the file size on the disk does not change. Bmp files contain raw data of the image. The file size of a BMP can only be reduced by resizing the dimensions of the image. This is why the decompressed image that we display has the same file size as the original image. If we were to save it as a .jpeg or .png then we would get a significantly lower file size regardless of the quality. The quality of the produced decompressed image depends on the block size and quant table used in the JPEG algorithm. It was found that if we altered these values, we would get different qualities of the original image.

Figure 3 shows the console-based UI selected image3.bmp, compression mode selected, and block size and quality factor selected.

```
: Quit
=====LoadFile=====
   1 Image1.bmp
   2 Image2.bmp
   3 Image3.bmp
   4 kinect-dinosaur.bmp
Select file, (the number surrounded by [ ]3
     : Compress
     : Display Image
     : Display File Size
     : Load Another Image
     : Quit
==Compression-Menu==
     : 8x8
Block size selected: (8, 8)
==Compression-Menu==
Enter Quality Factor (0 - 10) (0 means no compression, no data loss) (10 means almost max compression lots of data loss)
```

Figure 4 shows the console-based UI displaying metrics of the image after decompression

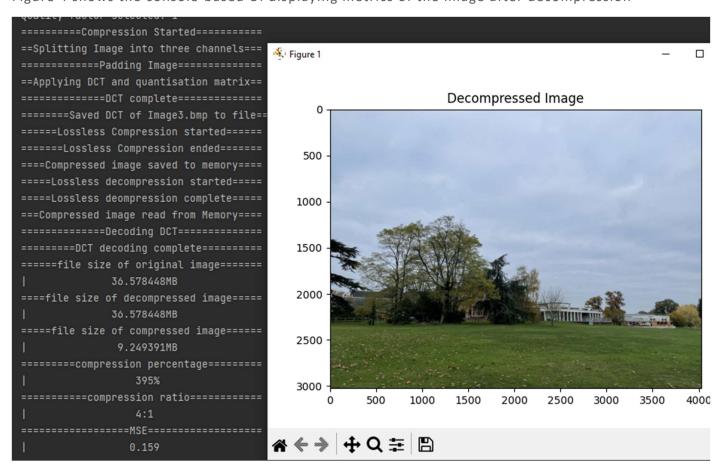


Figure 4 shows the UI displaying when each process of compression/ decompression starts and finishes during execution. The metrics are also displayed, including the MSE, compression ratio, compressed file size and original file size. The MSE is 0.159 which implies that the image is extremely similar to the original image in terms of its shapes and pixel arrangements. A higher MSE implies that the image's quality has changed from the original, it is most likely that in this case, the image's quality will seem much less. We successfully reduced the file size of this image to around 9.24MB in compressed memory from 36.57MB. This gave us a compression ratio of around 4:1 with a compression percentage of 395%. The image displayed is of the decompressed image, which has the same file size as the original image since it is saved as a BMP. If we were to save the image as a .jpeg or .png file, then we would see the exact same image as produced above.





Figure 6 Segmentation of area of interest in image



When deciding to see where we can detect changes in the image when applying compression, we decided to focus on a specific point in the image. The hollow red box applied to the image in figure (X)

Figure 7 Image3_recompressed_10_8x8.bmp Snipped Section

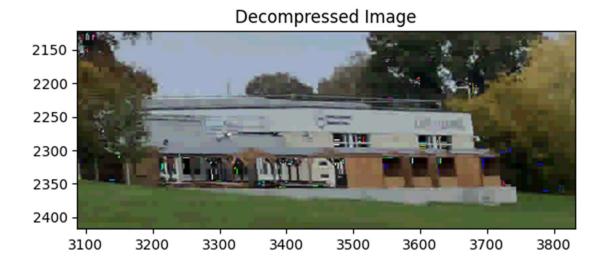


Figure 8. Image3_recompressed_1_8x8.bmp Snipped Section

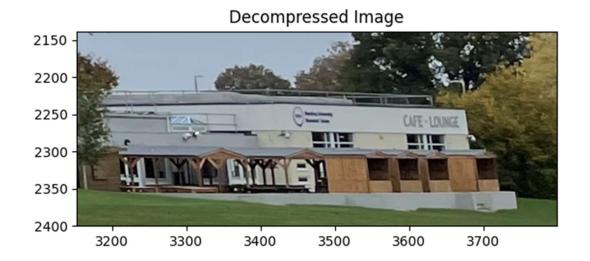


Image	MSE	Compression Ratio	Compression %	File size
Image3_recompressed_1_8x8.bmp	0.159	4:1	395%	9.249391MB
Image3_recompressed_10_8x8.bmp	190.476	101:1	10069%	0.363286MB

Figures 7 and 8 have been compressed using a pixel block size of 8x8 however Figure 7 uses a quality factor of 10 whereas Figure 8 uses a quality factor of 1. The quality factor defines the values inside the quantization table. The higher the quality factor the higher the values inside the table which will cause quantization to get rid of more DCT coefficients, reducing the data of the image. With a higher loss in data means less memory is occupied, however at the cost of having the quality of the image heavily reduced. This is evidenced in the figures above, where the compression with a quality factor of 10 reduces the size of the image by a ratio of 101:1, in comparison to the original size, but at the cost of having a much higher MSE (as shown by the pixel tearing and blur), whereas the compression with a quality factor of 1 achieves the opposite.

Similarly, the same process was applied to the other 3 images, the images for these are in the appendix of this report.

Conclusion

For this coursework we have developed an application for compressing, decompressing images, and displaying their metrics on a console GUI. The compression algorithm used by the application works by applying pixel blocks, DCT, Quantization, Zigzag method, the LZ77 and the Huffman Encoding lossless algorithms from the zlib compression software library to the images, then decompressing them by using the inverse of each method.

We have evidenced how loss-compression trade-off impacts the resultant quality and file size of the compressed images by applying pixel blocks of different sizes and quantization tables of different values. We have shown that higher values in the quantization table result in a bigger file size reduction at the cost of quality by calculating the MSE between compressed and original images.

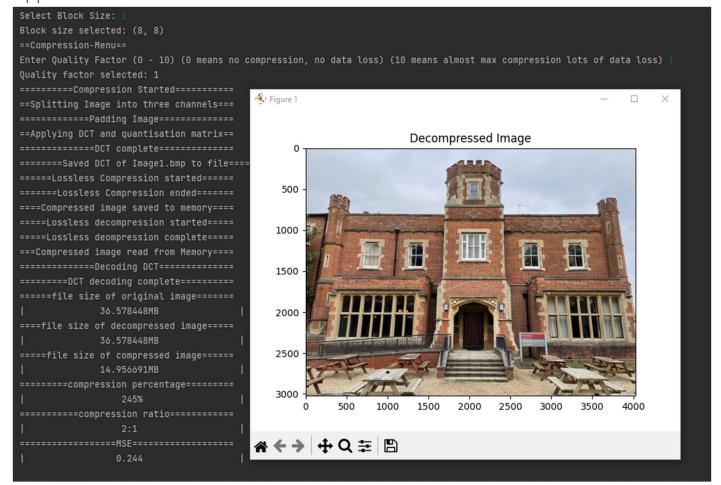
Based on our testing and design of the compression algorithm, the best result in terms of memory and quality of the images is achieved when we apply a quality factor of 1 and pixel blocks of size 8x8. Any quality factor higher than 1 will result in a higher size reduction, however at the cost of the image being exponentially worse. In theory, a pixel block size higher than 8x8 will increase the quality of the resultant image however the calculation takes a long time, making it hard to measure.

If we could further improve our findings of compression and decompression, we would implement other lossless compression algorithms, such as Arithmetic, Entropy or bit-plane coding. By implementing these we would have been able to compare which algorithm performs faster and more efficiently than the already lossless algorithms we have implemented in our design. This could have resulted in a much greater reduction in file size when our image is compressed.

References/Acknowledgements

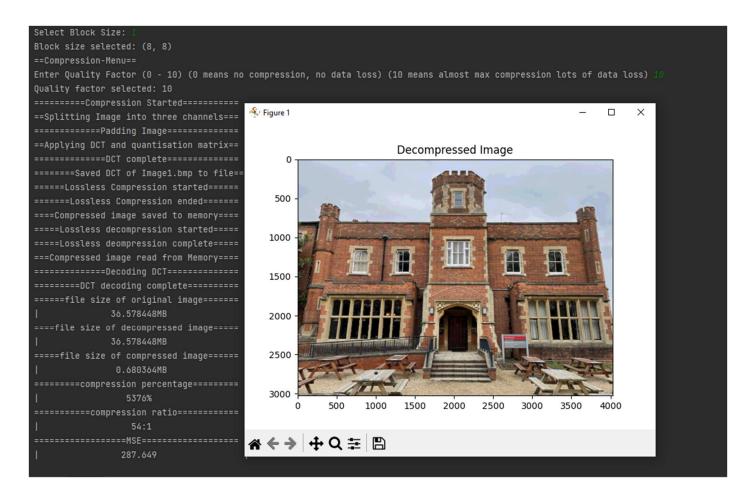
- [1] WhatIs.com. (n.d.). What is image compression and how does it work? [online] Available at: https://www.techtarget.com/whatis/definition/image-compression#:~:text=Image%20compression%20is%20a%20process.
- [2] dictionary.cambridge.org. (n.d.). COMPRESSION | meaning in the Cambridge English Dictionary. [online] Available at: https://dictionary.cambridge.org/dictionary/english/compression.
- [3] Stack Overflow. (n.d.). Why JPEG compression processes image by 8x8 blocks? [online] Available at: https://stackoverflow.com/questions/10780425/why-jpeg-compression-processes-image-by-8x8-blocks [Accessed 6 Dec. 2022].
- [4] www.image-engineering.de. (n.d.). How does the JPEG compression work? [online] Available at: https://www.image-engineering.de/library/technotes/745-how-does-the-jpeg-compression-work#:~:text=The%20JPEG%20compression%20is%20a.
- [5] Al-Ani, Muzhir & Awad, Fouad. (2013). THE JPEG IMAGE COMPRESSION ALGORITHM. International Journal of Advances in Engineering & Technology. 6. 1055-1062. https://www.researchgate.net/publication/268523100_THE_JPEG_IMAGE_COMPRESSION_ALGORITHM
- [6] Wikipedia. (2022). zlib. [online] Available at: https://en.wikipedia.org/wiki/Zlib [Accessed 6 Dec. 2022].
- [7] Wikipedia Contributors (2019). DEFLATE. [online] Wikipedia. Available at: https://en.wikipedia.org/wiki/DEFLATE.
- [8] Wikipedia. (2021). LZ77 and LZ78. [online] Available at: https://en.wikipedia.org/wiki/LZ77 and LZ78.
- [9] Dubey, S. (2022). compression-DCT. [online] GitHub. Available at: https://github.com/getsanjeev/compression-DCT [Accessed 6 Dec. 2022].

Appendix



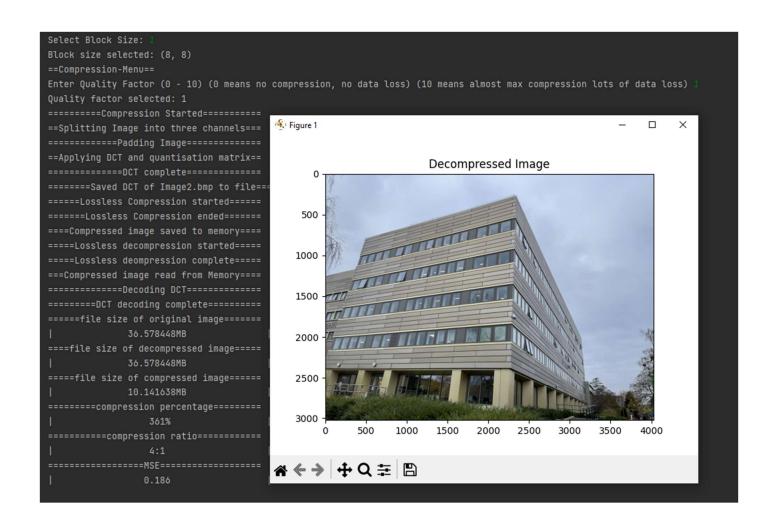
Decompressed Image

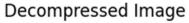




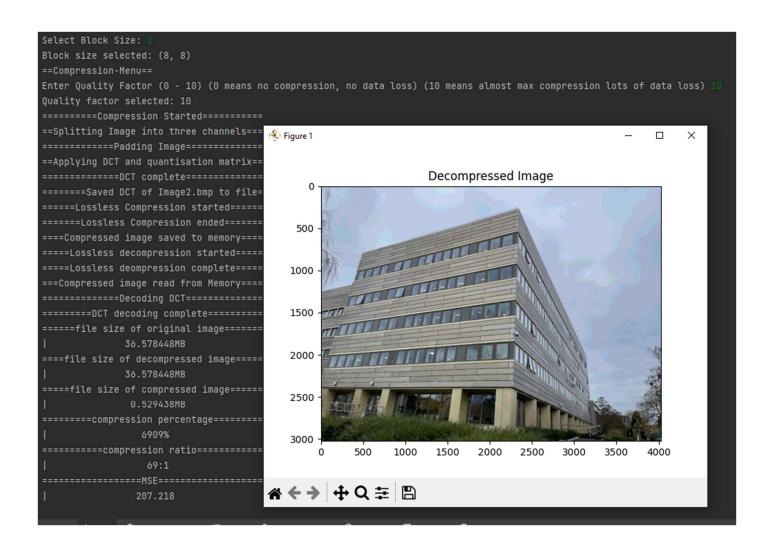
Decompressed Image



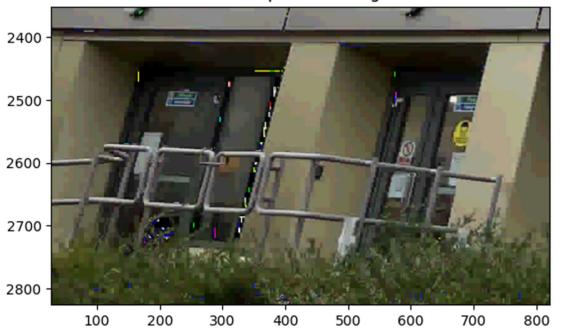


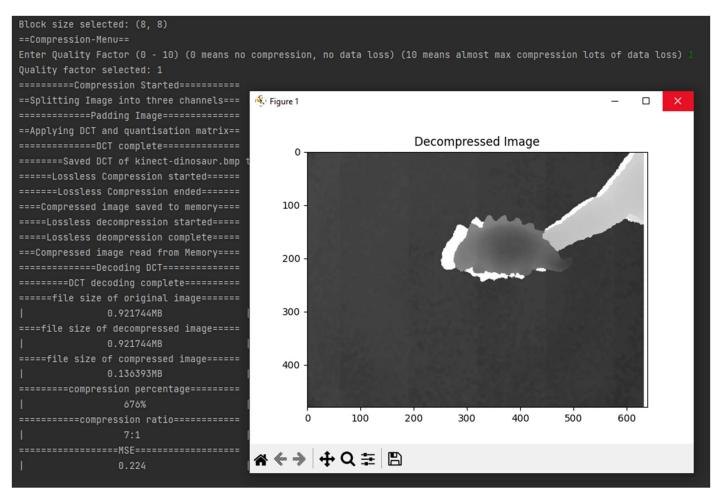


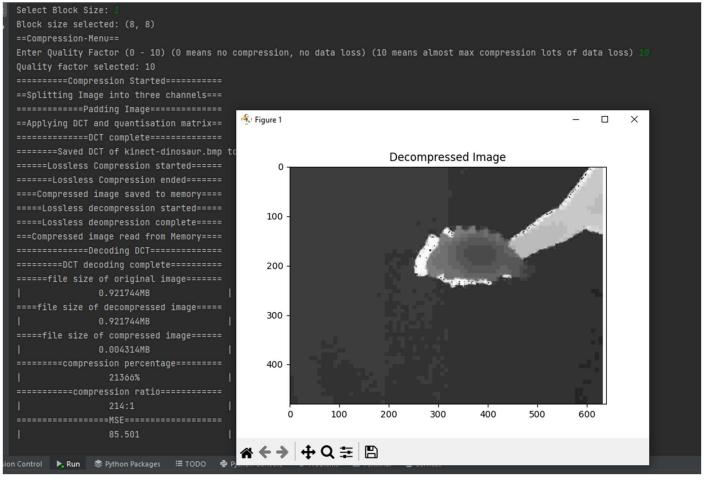












Original Image	Andrealit y Factor	Block size	Compresse d image size	MSE	Comp ratio	C o m p	Decompressed Image
0 1000 1300 2000 2000 3000 0 500 1000 1500 2000 2500 3000 3510 4000	1	8x8	14.95MB	0.24	245%	2: 1	Decompressed mage 500 3000 3000 2000 2000 2000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000
0 1000 13500 2000 2500 5 300 1000 1300 2000 2500 3000 3510 4500	1	4x4	15.273	0.24	239%	2: 1	Decompressed image 500 1300 2000 500 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000 1000
0 1000 13500 2000 27500 5 500 1500 1500 2000 2500 5000 1510 4000	5	4x4	2.86MB	55.9 76	1279%	1 3: 1	Decorpressed image 500 1300 2500 2500 500 1500 1500 2500 3000 500 1500 1500 2500 3000 3000 3000 3000 3000 4000
3000 a 500 1000 1500 2000 2500 3000 3510 4000	5	8x8	2.46MB	51.4 09	1490%	1 5: 1	Decompressed Image 500 1000 1300 2000 2000 3000 500 1500 1500 2000 2300 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000 3000
3000 a 500 1000 1500 2000 2500 3000 3500 4800	5	16x16	2.31MB	52.8 76	1582%	1 6: 1	Decompressed Image 1000 1000 1300 2000 3000 500 1000 1300 2000 2300 3000 3000 3000 4000

Code

Compression.py

```
import sys
import time

import cv2 as cv
from matplotlib import pyplot as plt
import os
import numpy as np
import argparse
```

```
import numpy as np
from scipy import fftpack
from huffman import HuffmanTree
import numpy as np
bSize = 8
def display files():
    global chosen file
```

```
def loadFile(string):
def Encoding Quantitisation Matrix(orig, quant):
def Decoding_Quantitsation_Matrix(orig, quant):
        :new shape[0],
   [height, width] = Green.shape # gets height and width of image, blue green or red
   boxHeight = np.int32(boxHeight)
   boxWidth = np.int32(boxWidth)
```

```
rowTwo = rowOne + bSize
            colOne = z * bSize
            BluePadded = PaddedImageBlue[rowOne: rowTwo, colOne: colTwo]
            GreenPadded = PaddedImageGreen[rowOne: rowTwo, colOne: colTwo]
            BlueDCT = cv2.dct(BluePadded)
            RedDCT = cv2.dct(RedPadded)
            RedReshaping = np.reshape(RedReordering, (bSize, bSize))
PaddedImageRed])
    cv2.imwrite('images with dct/{}encoded.bmp'.format(name), DCTtransformOfImage)
np.uint8(DCTtransformOfImage))
    time.sleep(2)
```

```
return fftpack.idctn(orig, axes=[1,3], norm='ortho'
def encoding compressed image(x):
    return zlib.compress(x.astype(np.int8).tobytes())
def decoding compressed image(orig, shape):
def compress(blocksize, quality):
   blocks.append(blocksize)
           decz = decoding compressed image(encz, encq.shape)
+" recompressed quant {} block {}x{}.bmp".format(qscale, bx, by), dec.astype(np.uint8))
            cv2.imwrite(
               dec.astype(np.uint8))
           cv2.destroyAllWindows()
            compression ratio =
```

```
def CompressMenu():
           CompressMenu()
```

Huffman.py

```
q = PriorityQueue()
 calc freq(self, arr):
```

zigzag.py [9] This is a third party source, we did not create and or claim ownership to this source code.

```
import numpy as np

def zigzag(imageInput):
    height = 0
    Vertex = 0
    vertexMinimum = 0
    heightMinimum = 0
    vertexMaximum = imageInput.shape[0]
    HeightMaximum = imageInput.shape[1]
    i = 0
```

```
if Vertex == vertexMinimum:
```

```
brea
```

```
output[v, h] = input[i]
    if v == vmax - 1:
        h = h + 1
    else:
        v = v + 1
    i = i + 1
    elif v < vmax - 1 and h > hmin:

    output[v, h] = input[i]
    v = v + 1
    h = h - 1
    i = i + 1

if v == vmax - 1 and h == hmax - 1:

    output[v, h] = input[i]
    break

return output
```