

DEPARTMENT OF ELECTRONIC AND TELECOMMUNICATION
UNIVERSITY OF MORATUWA

EN3551 DIGITAL SIGNAL PROCESSING

This is offered as a "EN3551 Digital Signal Processing" module's partial completion.



Assignment 02 : Application of 2D-DCT for Image Compression

200686J : Vishagar A.

7th of October, 2023

Abstract

*This report deals with the explanation of the solutions for the given questions in the assignment 02 of the EN3551 module. The solutions are explained in a way that it is easy to understand and follow. The solutions are explained with the help of the code snippets and the results. We are mainly focussed on some real world problems like "Application of 2D-DCT for Image Compression". **And I have used python as the programming language to find the solutions for the given problems.***

Contents

1	Introduction	3
2	Dataset	3
3	Methadology	3
3.1	Extracting and Preparing the Data	3
3.2	Implementation of DCT	3
3.3	Decompression and Visualization	3
4	Implementation and Results	3
4.1	Image 01 : camera256	3
4.2	Image 02 : boat512	4
4.3	Image 03 : peppers512	4
4.4	Image 01 : Camera256	5
5	Difficulty in Compression	5
6	References	5
7	Github Repository	5
8	Appendix	6
8.1	Extracting and Preparing the Data	6
8.2	DCT and Quantization	6
8.3	Coding Schemes	6
8.4	Inverse DCT and re-construction	7
8.5	Evaluation Metrics	7

1 Introduction

In this assignment, we are mainly focussed on a real world problem, **Image Compression** and we need to apply the 2D-DCT for the given image. And we need to find the compression ratio, percentage of zero and the PSNR value for the given images.

2 Dataset

The provided dataset contains **.mat** files of a image. We need to extract the necessary image data from the **.mat** file and use it for the further processing. I have assigned to use 3 images from the provided dataset and one image as per my wish.

3 Methodology

3.1 Extracting and Preparing the Data

As I mentioned earlier, we need to import the required **.mat** file and extract the image data from the **.mat** file using the relevant key name. Then if we wish we can visualize and see the imported image.

Then after importing the necessary image data, image array was divided in to blocks in a size of 8x8. And meanwhile as we are dealing with negative integer values also, I converted the default image data format from **np.uint8** to **np.int16**.

3.2 Implementation of DCT

Then after converting into blocks for better results the image data range was converted from **0 to 255** to **-128 to 127**, by subtracting 128 from each entry of the image matrix.

Then for each blocks which we have divided earlier, we need to apply the 2D-DCT. For that I have used the **dctn()** function in the **numpy** library.

And then the DCT coefficients were quantized using the given quantization matrix. Standard quantization matrix was given for a quality factor of 50 % and for the other factors the same matrix was multiplied by a scaling factor where the scaling factor can be derived using the given formula.

From each image block each individual element from the block was divided in to relevant element in the quantization matrix. And the derived matrix is the quantized matrix (named as S here). Then the quantized coefficients were encoded using the **zigzag scan** and the **run length encoding** methods.

3.3 Decompression and Visualization

After coding for visualization coded matrix will be again multiplied by the quantization matrix (Each element in each block will be multiplied with relevant entry in the quantization matrix in bitwise.) Then we apply inverse DCT using **idctn()** function. Then afterwards already deducted DC value will be compensated by adding the same DC value again. And by combining the small reconstructed blocks the image will be reconstructed. Then we can visualize the reconstructed image and compare it with the original image.

4 Implementation and Results

4.1 Image 01 : camera256

- Quality level 80

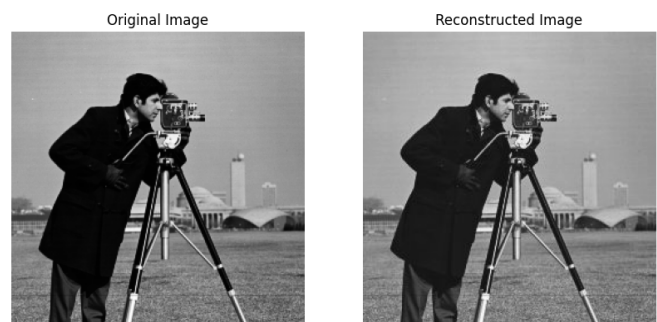


Figure 1: Provided sample file

- Percentage of zeroes : 74.942 %
- Peak SNR (in dB) : 35.7245 dB

- Quality level 35

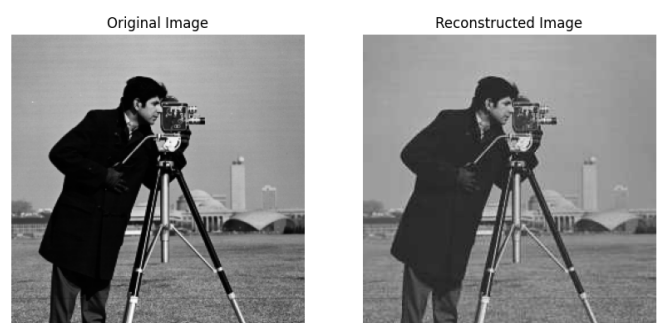


Figure 2: Provided sample file

- Percentage of zeroes : 88.275 %
- Peak SNR (in dB) : 30.3056 dB

- Quality level 15

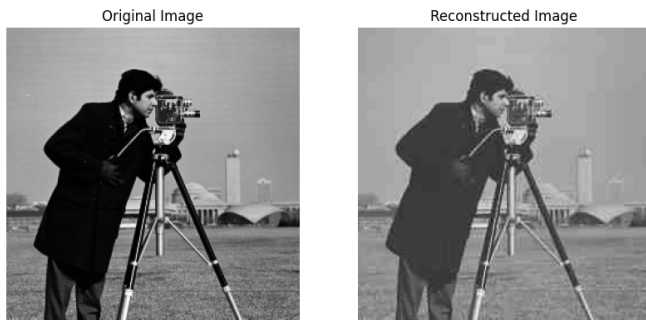


Figure 3: Provided sample file

- Percentage of zeroes : 93.348 %
- Peak SNR (in dB) : 27.536 dB

4.2 Image 02 : boat512

- Quality level 80



Figure 4: Provided sample file

- Percentage of zeroes : 78.343 %
- Peak SNR (in dB) : 38.0537 dB

- Quality level 35



Figure 5: Provided sample file

- Percentage of zeroes : 89.334 %
- Peak SNR (in dB) : 29.9607 dB

- Quality level 15



Figure 6: Provided sample file

- Percentage of zeroes : 93.820 %
- Peak SNR (in dB) : 60.1594 dB

4.3 Image 03 : peppers512

- Quality level 80



Figure 7: Provided sample file

- Percentage of zeroes : 79.283 %
- Peak SNR (in dB) : 36.9139 dB

- Quality level 35

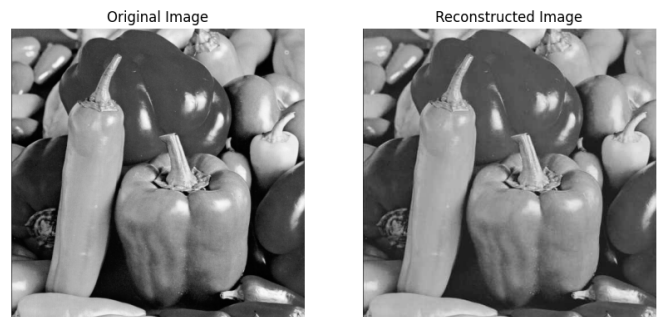


Figure 8: Provided sample file

- Percentage of zeroes : 91.356 %
- Peak SNR (in dB) : 33.9170 dB

- Quality level 15



Figure 9: Provided sample file

- Percentage of zeroes : 94.886 %
- Peak SNR (in dB) : 31.5075 dB

4.4 Image 01 : Camera256

- Quality level 80

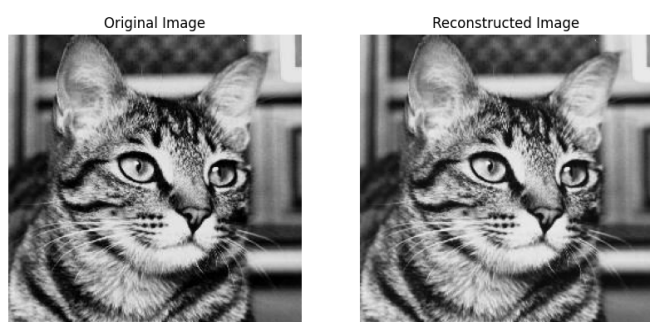


Figure 10: Provided sample file

- Percentage of zeroes : 67.567 %
- Peak SNR (in dB) : 40.7102 dB

- Quality level 35

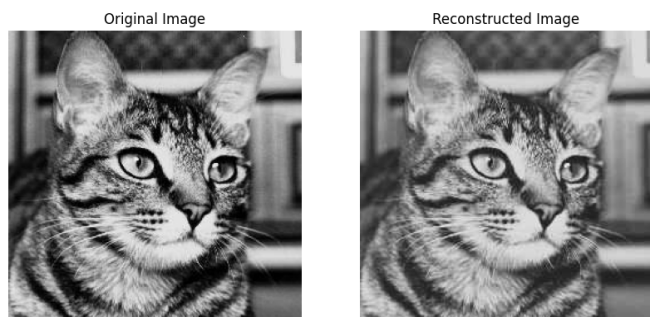


Figure 11: Provided sample file

- Percentage of zeroes : 82.170 %
- Peak SNR (in dB) : 28.4592 dB

- Quality level 15

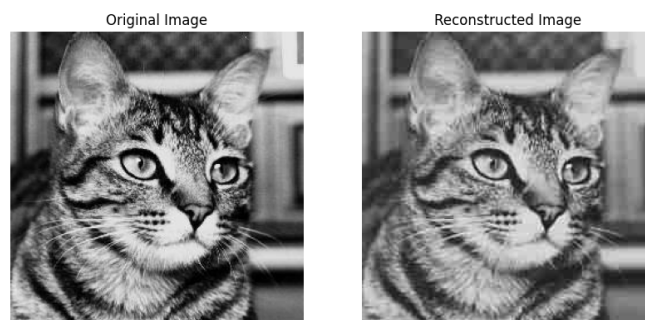


Figure 12: Provided sample file

- Percentage of zeroes : 89.554 %
- Peak SNR (in dB) : 25.7649 dB

5 Difficulty in Compression

6 References

- [Numpy Documentation](#)
- [Preferred Text Book](#)

7 Github Repository

Following is the link to my Github repository for this assignment.

[Github/EN3551_Assignment_02](#)

8 Appendix

8.1 Extracting and Preparing the Data

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 import scipy
4 import sympy
5
6 # Load the .mat file into a dictionary
7 mat_data = scipy.io.loadmat('SampleImages/
    camera256.mat')
8
9 # Extract the image data from the
    dictionary
10 image_data = mat_data['camera256']
11
12 image_data = image_data.astype(np.int16)
13
14 # Display the image using Matplotlib
15 plt.imshow(image_data, cmap='gray')
16 plt.axis('on')
17 plt.show()
18
19 # Separate image into 8x8 blocks
20 block_size = 8
21 blocks = []
22 for i in range(0, image_data.shape[0],
    block_size):
23     for j in range(0, image_data.shape[1],
        block_size):
24         blocks.append(image_data[i:i+
            block_size, j:j+block_size])
25
26 # Convert blocks to numpy array
27 blocks = np.array(blocks)
28 blocks = blocks - 128

```

8.2 DCT and Quantization

```

1 #Apply 2 dimensional DCT to each block
2
3 dct_blocks = []
4 for i in range(blocks.shape[0]):
5     dct_blocks.append(scipy.fftpack.dctn(
        blocks[i], norm='ortho'))
6 dct_blocks = np.array(dct_blocks)
7
8 # Quantize the DCT coefficients
9
10 quantization_matrix = np.array([[16, 11,
    10, 16, 24, 40, 51, 61],[12, 12, 14, 19,
    26, 58, 60, 55],[14, 13, 16, 24, 40,
    57, 69, 56],[14, 17, 22, 29, 51, 87, 80,
    62],[18, 22, 37, 56, 68, 109, 103,
    77],[24, 35, 55, 64, 81, 104, 113,
    92],[49, 64, 78, 87, 103, 121, 120,
    101],[72, 92, 95, 98, 112, 100, 103,
    99]])
11
12 quality_factor = 5
13
14 if quality_factor > 50:
15     scale = (100 - quality_factor) / 50
16     quantization_matrix =
        quantization_matrix * scale
17 else:

```

```

18     scale = quality_factor / 50
19     quantization_matrix =
        quantization_matrix * scale
20
21 quantized_dct_blocks = []
22
23
24 for x in dct_blocks:
25     s = np.zeros((8, 8))
26     for i in range(8):
27         for j in range(8):
28             s[i][j] = round(x[i][j] /
                quantization_matrix[i][j])
29     quantized_dct_blocks.append(s)
30
31 quantized_dct_blocks = np.array(
    quantized_dct_blocks)

```

8.3 Coding Schemes

```

1 # Coding the DC coefficients
2
3 dc_coefficients = []
4 for x in quantized_dct_blocks:
5     dc_coefficients.append(x[0][0])
6
7 # differential pulse code modulation for DC
    coefficients
8
9 dpcm_dc_coefficients = []
10 dpcm_dc_coefficients.append(dc_coefficients
    [0])
11 for i in range(1, len(dc_coefficients)):
12     dpcm_dc_coefficients.append(
        dc_coefficients[i] - dc_coefficients
        [i-1])
13
14 # run,level modulation for AC pairs
15
16 run_level_pairs_blocks = []
17
18 for x in quantized_dct_blocks:
19     run = 0
20     level = 0
21     run_level_pairs = []
22     for i in range(1, 8):
23         for j in range(1, 8):
24             if x[i][j] != 0:
25                 run_level_pairs.append((run
                    , level))
26                 run = 0
27                 level = x[i][j]
28             else:
29                 run += 1
30
31     run_level_pairs_blocks.append(
        run_level_pairs)
32
33 # Entropy coding for run-level pairs
34
35 entropy_coded_blocks = []
36
37 for x in run_level_pairs_blocks:
38     entropy_coded_block = []
39     for i in range(len(x)):
40         if i == 0:
41             entropy_coded_block.append((x[i]

```

```

    ][0], x[i][1]))
42     else:
43         if x[i][0] == 0 and x[i][1] == 0:
44             entropy_coded_block.append
45                 ((15, 0))
46         else:
47             entropy_coded_block.append
48                 ((x[i][0], x[i][1]))
49     entropy_coded_blocks.append(
50         entropy_coded_block)

```

8.4 Inverse DCT and re-construction

```

1 # inverse quantization
2 inverse_quantized_dct_blocks = []
3
4 for x in quantized_dct_blocks:
5     s = np.zeros((8, 8))
6     for i in range(8):
7         for j in range(8):
8             s[i][j] = x[i][j] *
9                 quantization_matrix[i][j]
10    inverse_quantized_dct_blocks.append(s)
11 inverse_quantized_dct_blocks = np.array(
12     inverse_quantized_dct_blocks)
13 # perform inverse DCT
14
15 inverse_dct_blocks = []
16
17 for x in inverse_quantized_dct_blocks:
18     inverse_dct_blocks.append(scipy.fftpack
19         .idctn(x, norm='ortho'))
20 inverse_dct_blocks = np.array(
21     inverse_dct_blocks)
22
23 inverse_dct_blocks = inverse_dct_blocks +
24     128
25 inverse_dct_blocks = inverse_dct_blocks.
26     astype(int)
27
28 num_blocks = int(np.sqrt(inverse_dct_blocks
29     .shape[0]))
30 block_size = inverse_dct_blocks.shape[1]
31
32 # Reshape the blocks array
33 blocks_resaped = inverse_dct_blocks.
34     reshape((num_blocks, num_blocks,
35         block_size, block_size))
36
37 # Initialize an empty image
38 reconstructed_image = np.zeros((num_blocks
39     * block_size, num_blocks * block_size))
40
41 # Reconstruct the image from blocks
42 for i in range(num_blocks):
43     for j in range(num_blocks):
44         reconstructed_image[i*block_size:(i
45             +1)*block_size, j*block_size:(j
46                 +1)*block_size] =
47             blocks_resaped[i, j, :, :]
48
49 # Visualize the images
50 fig, ax = plt.subplots(1, 2)
51 ax[0].imshow(image_data, cmap='gray')

```

```

41 ax[0].set_title('Original Image')
42 ax[0].axis('off')
43 ax[1].imshow(reconstructed_image, cmap='
44     gray')
45 ax[1].set_title('Reconstructed Image')
46 ax[1].axis('off')
47 plt.show()

```

8.5 Evaluation Metrics

```

1 # find percentage of zeroes
2 num_zeroes = 0
3 for x in quantized_dct_blocks:
4     for i in range(8):
5         for j in range(8):
6             if x[i][j] == 0:
7                 num_zeroes += 1
8
9 percentage_zeroes = num_zeroes / (
10     quantized_dct_blocks.shape[0] * 64) *
11     100
12 print("Percentage of Zeroes : ",
13     percentage_zeroes)
14
15 # two dimensional mean square error
16 error_matrix = image_data -
17     reconstructed_image
18 mse = np.mean(np.square(error_matrix))
19 print("Mean Square Error : ",mse)
20
21 # calculate the peak signal to noise ratio
22 psnr = 20 * np.log10((255) / mse)
23 print("Peak Signal to Noise Ratio (PSNR) :
24     ",psnr)

```