

Department of Electronic & Telecommunication Engineering University of Moratuwa

EN3551 - DIGITAL SIGNAL PROCESSING

APPLICATION OF 2D-DCT FOR IMAGE COMPRESSION

MANIMOHAN T. 200377M

ABSTRACT

This assignment report focuses on the application of 2D-DCT for image compression, where the Discrete Cosine Transform is used to compress digital images. The report provides a theoretical background on 2D-DCT, explaining its properties, orthogonality, and energy preservation. It also discusses the quantization and coding process for compression and the subsequent decompression steps.

The assignment involves the compression of various images with different quality levels. The performance is evaluated in terms of the percentage of zeros, peak signal-to-noise ratio (PSNR), and visual quality. Through this assignment, students gain a practical understanding of the principles of image compression using 2D-DCT and the impact of different quality levels on compression results. This practical experience enhances their knowledge of digital signal processing and image compression techniques.

1 DETECTING HARMONICS IN NOISY DATA AND SIGNAL INTERPOLATION USING DFT

1.1 Procedure

1.1.1 2-D DCT - Application to Image Compression

Suppose the image to be compressed has size $M \times N$, where M and N are multiples of 8 and represented with 8 bits. The image is first divided into 8×8 blocks. The steps described below are applied to each of these blocks. Let B be a representative 8×8 data block.

Step 1: Level-Off and 2-D DCT

Since DCT gives a lower DC coefficient when pixel values are in the range -128 to 127 rather than in the range 0 to 255, B is leveled off by subtracting 128 from each entry. For illustration purposes, we take the (16, 18)-th block of the 256×256 image Lena as matrix B.

$$B = \begin{bmatrix} 125 & 134 & 137 & 139 & 138 & 138 & 141 & 142 \\ 113 & 119 & 126 & 134 & 139 & 141 & 144 & 149 \\ 80 & 95 & 103 & 106 & 114 & 127 & 141 & 147 \\ 63 & 65 & 53 & 62 & 75 & 86 & 108 & 130 \\ 93 & 80 & 60 & 33 & 35 & 35 & 52 & 69 \\ 126 & 108 & 88 & 74 & 53 & 45 & 35 & 32 \\ 130 & 116 & 90 & 96 & 62 & 63 & 55 & 49 \\ 115 & 80 & 61 & 65 & 68 & 88 & 68 & 75 \end{bmatrix}$$

The modified data block after subtracting 128 from B is given by,

$$\tilde{B} = \begin{bmatrix} -3 & 6 & 9 & 11 & 10 & 10 & 13 & 14 \\ -15 & -9 & -2 & 6 & 11 & 13 & 16 & 21 \\ -48 & -33 & -25 & -22 & -14 & -1 & 13 & 19 \\ -65 & -63 & -75 & -66 & -53 & -42 & -20 & 2 \\ -35 & -48 & -68 & -95 & -93 & -93 & -76 & -59 \\ -2 & -20 & -40 & -54 & -75 & -83 & -93 & -96 \\ 2 & -12 & -38 & -32 & -66 & -65 & -73 & -79 \\ -13 & -48 & -67 & -63 & -60 & -40 & -60 & -53 \end{bmatrix}$$

Applying 2-D DCT to \tilde{B} , we obtain,

$$C = T_8 B \tilde{T}_8 = \begin{bmatrix} -272.3750 & 17.1771 & 46.7784 & 4.2270 & 6.1250 & -0.5799 & 0.2421 & -8.4813 \\ 182.5146 & -109.5361 & -28.0398 & -25.1231 & -8.9567 & -7.0036 & -6.1703 & 10.3445 \\ 117.4897 & 19.1429 & -30.8911 & 15.7065 & 1.7309 & 6.7882 & 5.4116 & -8.5788 \\ -23.4612 & 97.7162 & -0.0872 & -7.0795 & -1.0461 & 2.6980 & -2.9351 & 2.5446 \\ -48.3750 & -35.2958 & 27.0407 & 6.4060 & 4.1250 & -7.3615 & 3.5470 & 4.3781 \\ 15.8386 & -7.1745 & -7.9234 & -6.7518 & -0.5166 & 3.8019 & -8.3849 & -7.4654 \\ 0.4477 & 1.3348 & -5.5884 & 3.4787 & -4.6404 & -0.7361 & 4.6411 & 2.6707 \\ -3.6673 & 9.8948 & 6.2377 & -7.3148 & -7.0342 & 1.0065 & -0.6727 & 2.3138 \end{bmatrix}$$

Alternatively, one may apply MATLAB function dct2 to \tilde{B} to obtain C.

Step 2: Quantization

A critical step in image compression is quantization and coding of the DCT coefficients in C. The quantization may be achieved by converting matrix $C = \{c_{i,j}\}$ to matrix $S = \{s_{i,j}\}$ determined by,

$$s_{i,j} = \text{round}\left(\frac{c_{i,j}}{q_{i,j}}\right)$$

where $Q = \{q_{i,j}\}$ is a quantization matrix that may be selected according to a desired quality level of the compression. In the JPEG standard, subjective experiments involving the human visual system have produced their own quantization matrices. For example, with a quality level 50 (on a 1-to-100 scale) we have,

$$Q_{50} = \begin{bmatrix} 16 & 11 & 10 & 16 & 24 & 40 & 51 & 61 \\ 12 & 12 & 14 & 19 & 26 & 58 & 60 & 55 \\ 14 & 13 & 16 & 24 & 40 & 57 & 69 & 56 \\ 14 & 17 & 22 & 29 & 51 & 87 & 80 & 62 \\ 18 & 22 & 37 & 56 & 68 & 109 & 103 & 77 \\ 24 & 35 & 55 & 64 & 81 & 104 & 113 & 92 \\ 49 & 64 & 78 & 87 & 103 & 121 & 120 & 101 \\ 72 & 92 & 95 & 98 & 112 & 100 & 103 & 99 \end{bmatrix}$$

Quantization matrices of other quality levels can be obtained by multiplying the above standard quantization matrix by a scaling factor τ where,

$$\tau = \begin{cases} \frac{100 - \text{quality level}}{50}, & \text{quality level} > 50\\ \frac{50}{\text{quality level}}, & \text{quality level} < 50 \end{cases}$$

The scaled quantization matrix is then rounded and clipped to integer values between 0 and 255. With C obtained in Step 1 and $Q = Q_{50}$, the matrix S in Eq. 10 is computed as

Evidently, if a quantization matrix with a reduced quality level is used, S will contain more zeros, thus a higher compression can be achieved at the cost of reduced image quality.

Step 3: Coding

The quantized matrix S is then converted by an encoder to a stream of binary data like $\{1001101,\ldots\}$. Detailed coverage of the coding process is beyond the scope of this experiment. Instead, we mention the following.

- * The DC coefficients of the 8×8 blocks are encoded by a differential pulse-code modulation (DPCM) coding (also referred to as DC prediction).
- * Because the location of the retained coefficients varies from block to block, the quantized AC coefficients are zigzag scanned and ordered into (run, level) pairs, where "level" means the value of a nonzero coefficient, and "run" means the number of zero coefficients preceding it.
- * These (run, level) pairs are entropy coded, that is, longer codes for less frequent pairs and vice versa.

Step 4: Decompression

This is a step to be carried out at the receiver to reconstruct the image. It consists of the following three simple operations.

* Pointwise multiplication of matrix S with the quantization matrix Q to get an image block $R = Q \circ S$. In our example, the Q matrix was taken to be Q_{50} that yields,

* Apply 2-D inverse DCT to matrix R. Note that this can be done by using the MATLAB command idct2. Alternatively, it can be done using $D_8^T R D_8$. This gives

$$E = D_8^T R D_8 =$$

2.71	181	2.1228	2.2693	4.2600	7.6484	10.8548	12.7203	13.3406
-18.1	1034	-10.6760	-0.3337	7.7581	11.5839	12.5496	12.8140	13.1153
-43.5	2294	-34.5413	-23.6264	-15.8552	-9.7946	-0.4317	12.7343	23.0141
-57.9	9809	-58.9924	-63.0283	-67.8737	-64.7907	-46.4675	-17.7381	4.4189
-40.7	7650	-50.9610	-69.2674	-88.8143	-98.8380	-92.2600	-73.6202	-57.6914
0.10	073	-12.7805	-34.0432	-57.0831	-76.3805	-89.4098	-96.5154	-99.3926
6.49	904	-8.6805	-28.5866	-43.2078	-52.0062	-61.1538	-73.3190	-83.0164
$\lfloor -25.5 \rfloor$	3942	-44.5619	-65.1582	-70.4424	-60.8203	-50.0267	-47.8854	-50.9752

* Finally, the effect of the "level-off" operation in Step 1 is taken into account by adding 128 to the entries of the matrix E obtained above. This yields the reconstructed image block that mimics the image block B as,

$$B = E + 128 =$$

130.7181	130.1228	130.2693	132.2600	135.6484	138.8548	140.7203	141.3406
109.8966	117.3240	127.6663	135.7581	139.5839	140.5496	140.8140	141.1153
84.7706	93.4587	104.3736	112.1448	117.2054	127.5683	139.7343	151.0141
70.0191	69.0076	64.9717	60.1263	63.2093	81.5325	110.2619	132.4189
87.2350	77.0390	58.7326	39.1857	29.1620	35.7400	54.3798	70.3086
128.1073	115.2195	93.9568	70.9069	51.6195	38.5902	31.4846	28.6074
134.4904	119.3195	99.4134	84.7922	75.9938	66.8462	54.6810	44.9836
102.6058	83.4381	62.8418	57.5576	67.1797	77.9733	80.1146	77.0248

On comparing the block B^{\sim} with block B (given in Step 1), we see B^{\sim} approximates B quite well. For an image of size $M \times N$, the number of blocks is $M/8 \times N/8$, and the steps described above are applied to each of these blocks for a selected quality level. We count the number of zeros contained in the S matrices (see Step 2) in order to compute the compression factor because the total number of zeros indicates how the image has been compressed. As expected, the use of reduced quality levels leads to a higher percentage of total zeros at the cost of a degraded image.

1.1.2 Peak Signal to Noise Ratio (PSNR)

The PSNR is defined as:

$$PSNR = 20 \log_{10} \left(\frac{\psi_{\text{max}}}{\sigma_e} \right)$$

where ψ_{max} denotes the maximum light intensity of the original image and σ_e^2 denotes the mean squared error. For an 8-bit digital image, $\psi_{\text{max}} = 255$. Furthermore, σ_e^2 can be computed in MATLAB applying the command mean twice (over the two dimensions) on a

squared error matrix. The error matrix is given by $I_{\text{rec}} - I_{\text{original}}$, where I_{rec} and I_{original} are the reconstructed and original images.

1.1.3 Tasks

1. Download 3 images assigned to your index number. Select one additional image of your choice, which is not in the provided set of images.

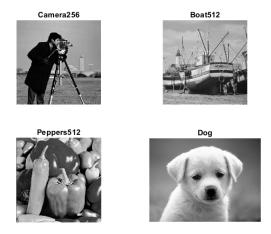


Figure 1 — Load the image

2. Compress each of these images using the DCT-based method described above with three quality levels. The quality levels are also assigned based on your index number.



Figure 2 — Compress and observe image 1 at quality level 60

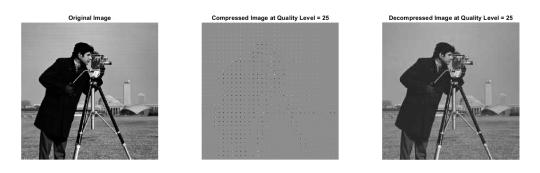


Figure 3 — Compress and observe image 1 at quality level 25



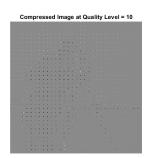




Figure 4 — Compress and observe image 1 at quality level 10



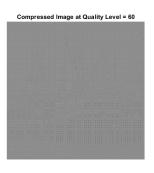




Figure 5 — Compress and observe image 2 at quality level 60







Figure 6 — Compress and observe image 2 at quality level 25





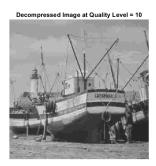
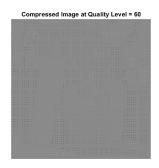


Figure 7 — Compress and observe image 2 at quality level 10





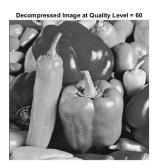


Figure 8 — Compress and observe image 3 at quality level 60

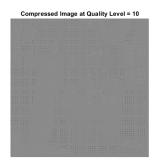






Figure 9 — Compress and observe image 3 at quality level 25





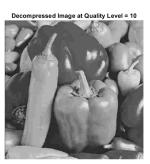


Figure 10 — Compress and observe image 3 at quality level 10



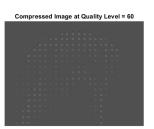




Figure 11 — Compress and observe image 4 at quality level 60



Figure 12 — Compress and observe image 4 at quality level 25



Figure 13 — Compress and observe image 4 at quality level 10

- 3. Observe the results in terms of:
- Percentage of zeros
- Quality in terms of peak signal-to-noise ratio (PSNR)
- Visual quality of the compressed images as related to the quality level.

```
When Quality Level = 60
Percentage of Zeros = 83.1192%
Peak Signal-to-Noise Ratio(PSNR) = 32.5619dB
When Quality Level = 25
Percentage of Zeros = 90.6219%
Peak Signal-to-Noise Ratio(PSNR) = 29.1581dB
When Quality Level = 10
Percentage of Zeros = 95.0058%
Peak Signal-to-Noise Ratio(PSNR) = 26.2428dB
```

Figure 14 — Compress and observe the facts of image 1

```
When Quality Level = 60
Percentage of Zeros = 85.1223%
Peak Signal-to-Noise Ratio(PSNR) = 41.5961dB
When Quality Level = 25
Percentage of Zeros = 91.3395%
Peak Signal-to-Noise Ratio(PSNR) = 38.2472dB
When Quality Level = 10
Percentage of Zeros = 95.3625%
Peak Signal-to-Noise Ratio(PSNR) = 34.8973dB
```

Figure 15 — Compress and observe the facts of image 2

```
When Quality Level = 60
Percentage of Zeros = 87.1059%
Peak Signal-to-Noise Ratio(PSNR) = 41.4033dB
When Quality Level = 25
Percentage of Zeros = 93.0378%
Peak Signal-to-Noise Ratio(PSNR) = 39.1156dB
When Quality Level = 10
Percentage of Zeros = 95.9991%
Peak Signal-to-Noise Ratio(PSNR) = 36.153dB
```

Figure 16 — Compress and observe the facts of image 3

Compress and observe image 4

```
When Quality Level = 60
Percentage of Zeros = 96.3632%
Peak Signal-to-Noise Ratio(PSNR) = 30.3488dB

When Quality Level = 25
Percentage of Zeros = 97.3148%
Peak Signal-to-Noise Ratio(PSNR) = 30.3305dB

When Quality Level = 10
Percentage of Zeros = 98.6434%
Peak Signal-to-Noise Ratio(PSNR) = 30.2499dB
```

Figure 17 — Compress and observe the facts of image 4

For all 4 images, the following 3 observations can be seen when the quality level decreases from 60 to 25 to 10

- a) An increase in the percentage of zeros can be observed when reducing the quality level. This phenomenon can be explained using the following equations. When the quality level is low, the scaling factor τ becomes large. For example, at quality level 90, τ is 0.2, whereas at quality level 10, τ becomes 5. A large τ results in a large quantization matrix Q, which, in turn, leads to a smaller scaled quantization matrix S. Consequently, when employing a quantization matrix Q with a lower quality level, the scaled quantization matrix S contains more zeros, achieving higher compression at the expense of reduced image quality. This expected behavior of the percentage of zeros can be observed in the experimental results obtained from all four images.
- b) A decrease in Peak Signal-to-Noise Ratio (PSNR) can be noted as the quality level decreases. This can be explained using the following equations. When the quality level decreases, the deviation between the reconstructed image and the original image increases. As a result, the mean squared error value increases, leading to a decrease in PSNR, as indicated by the equation. This anticipated behavior of PSNR is evident in the experimental results obtained from all four images.
- c) Visual quality experiences a reduction as the quality level decreases. However, the relationship between visual quality and the quality value doesn't appear to follow a linear

pattern. This can be elaborated as follows.

- The relationship between quality level and perceived visual quality is non-linear. Decreasing the quality level suggests increased compression and a potential decrease in visual quality. However, the extent of perceived quality reduction is influenced by various factors and may not strictly correlate with the numerical change in quality level. For instance, reducing the quality value from 60 to 25 doesn't necessarily result in an equivalent reduction in perceived visual quality. This non-linear relationship is evident in the results from Task 2, where it's observed that a decrease from 60 to 25 doesn't significantly degrade perceived visual quality.
- The threshold effect is noteworthy. At very high-quality levels, slight reductions in quality may not be noticeable to the human eye. However, as the quality level decreases, there's a point where even a minor reduction can lead to a significant loss in visual quality. This phenomenon is referred to as the "threshold effect."
- Diminishing Returns, As the quality level continues to decrease, visual quality degrades more rapidly. Beyond a certain point, further reductions may lead to a disproportionately large drop in visual quality.
- Visual quality is subjectively perceived and can vary from person to person.
 Some individuals may be more sensitive to compression artifacts, while others may be less discerning.
- 4. Some images are more difficult to compress than others in the sense that the visual (subjective) quality cannot be maintained as easily as others for a given compression ratio. Observe the compression results obtained in task (2) from this perspective and explain why?

Some images present a greater challenge in terms of compression while preserving visual (subjective) quality compared to others. When examining the results from Task 2, it's essential to understand the reasons behind this phenomenon.

The difficulty of effectively compressing an image while retaining visual quality is influenced by several key factors: image content, complexity, and frequency distribution. Images featuring intricate textures, patterns, and high-frequency details tend to be more demanding to compress without introducing noticeable artifacts.

Image Content: Natural images, such as landscapes and portraits, are characterized by intricate details, subtle color variations, and complex textures. This complexity makes it more challenging to achieve compression without compromising visual quality compared to artificial images like text or diagrams.

Image Complexity: Images with a high degree of randomness or entropy, such as those with considerable noise or fine textures, pose greater difficulties for effective compression.

The compression algorithm struggles to identify redundant patterns or predictable areas within these images.

Frequency Distribution: Images with a broad spectrum of frequencies, including high-frequency components like sharp edges and fine details, also present challenges in compression. High-frequency information is crucial for preserving image sharpness and clarity but is more susceptible to compression artifacts such as ringing and blurring.

In general, images with simpler content, lower complexity, and a narrower frequency distribution are more amenable to compression without compromising visual quality. For instance, images featuring large areas of uniform color, smooth gradients, or simple geometric shapes typically lend themselves well to compression.

These factors, as described, become evident upon reviewing the outcomes from Task 2.



Figure 18 — Compress and observe image 1 at quality level 25

Considering the image in question, it's evident that the quality of smooth regions like the person's coat and the sky remains relatively consistent. However, there is a noticeable loss of fine details in the grass's texture, and the fine lines of the tripod have become less distinct.



Figure 19 — Compress and observe image 2 at quality level 25

In the image above, the compression has had a more pronounced effect on fine details like the ship's masts and halyards, in contrast to the relatively smooth areas like the sky and the bottom of the ship.







Figure 20 — Compress and observe image 3 quality level 25

This image predominantly consists of fine textures and lacks substantial smooth regions. Consequently, it has suffered significant degradation as a result of compression. This degradation is particularly conspicuous at the edges of the peppers.







Figure 21 — Compress and observe image 4 quality level 25

In the image depicted above, there is a noticeable preservation of smooth regions, like the solid portions, during the compression process. However, the compression appears to have caused significant degradation in the sharpness of the edges. Additionally, numerous areas within the image containing intricate parts have experienced substantial deterioration as a result of the compression.

1.1.4 Useful MATLAB Commands

- plot: plots a function
- fft: calculates the DFT
- dct2: returns the two-dimensional discrete cosine transform of the input
- idct2: returns the two-dimensional inverse discrete cosine transform of the input
- load: load contents from a file into a MATLAB workspace.
- norm: returns the norm of a vector input.

1.1.5 References

- 1. Alan Oppenheim, Ronald Schafer, "Discrete Time Signal Processing," Prentice Hall Signal Processing (3rd edition), 2009.
- 2. S.A. Khayam, "The discrete cosine transform (DCT): Theory and Application," Report, Dept of ECE, Michigan State University, March 2003.
- 3. K. Cabeen and P. Gent, "Image compression and the discrete cosine transform," College of the Redwoods.
- 4. A related demo by Berkeley EECS: JPEG DCT Demo

2 APPENDIX

Github Link :- Assignment-02

2.1 Application of 2D-DCT for Image Compression

1. Download 3 images assigned to your index number. You have to select one more image of your choice, which is not in the provided set of images.

```
% Load the images
  load('camera256.mat');
                             % Replace with your image file names
  load('boat512.mat');
3
  load('peppers512.mat');
4
   image = imread("dog.jpg");
5
  % Display the images
6
   figure;
7
   subplot(2, 2, 1);
8
   imshow(camera256, []);
9
   title('Camera256');
10
11
   subplot(2, 2, 2);
12
   imshow(boat512, []);
13
   title('Boat512');
14
15
   subplot(2, 2, 3);
16
   imshow(peppers512, []);
17
   title('Peppers512');
18
19
   subplot(2, 2, 4);
20
   imshow(image, []);
21
   title('Dog');
```

Listing 2.1 — Load images

- 2. Compress each of these images by the DCT-based method described above with three quality levels. The quality levels are also assigned based on your index number.
- 3. Observe the results in terms of:
- Percentage of zeros
- Quality in terms of peak signal-to-noise ratio (PSNR)
- Visual quality of the compressed images as related to the quality level.

```
clc;
clear;
close all;
```

```
4
   % Loading the required image
5
   load("camera256.mat")
6
   image = camera256;
7
8
9
   [rows, cols, ~] = size(image);
10
11
   % Determine the number of 8x8 blocks needed along each dimension
12
   numBlocksRows = ceil(rows/8);
13
   numBlocksCols = ceil(cols/8);
14
15
   \% Make a copy of the original image before doing operations using it
16
   original_image = image;
17
18
   % Divide the image into 8x8 blocks
19
   image = mat2cell(image, 8 * ones(1, numBlocksRows), 8 * ones(1,
20
      numBlocksCols));
21
   % Subtract 128 from every single element
22
   image = cellfun(@(x) x - 128, image, 'UniformOutput', false);
23
24
   % Apply DCT-2 to each block
25
   image = cellfun(@dct2, image, 'UniformOutput', false);
26
27
   % Define the quality levels for compression.
28
   qualityLevels = [60, 25, 10];
29
   numQualityLevels = length(qualityLevels);
30
31
32
   % Loop through each quality level.
   for level = 1:numQualityLevels
33
34
       \% Calculate the quality factor for this level.
       qualitylevel = qualityLevels(level);
35
       % Defining Quantization Matrix
36
       quantizationMatrix = createJPEGQuantizationMatrix(qualitylevel);
37
38
       % Quantize each block
39
       quantizedmatrix = cellfun(@(x)
40
          quantizeDCTCoefficients(x,quantizationMatrix),image,'UniformOutputt'
       ,false);
41
42
       % Combine quantizedmatrix into a single image
43
       quantizedmatrix = cell2mat(quantizedmatrix);
44
       quantizedmatrix_copy = quantizedmatrix;
45
46
```

```
% Converting quantized matrix into a cell array
47
       quantizedmatrix =
48
          mat2cell(quantizedmatrix,8*ones(1,numBlocksRows),8*ones(1,
       numBlocksCols));
49
50
       % Encoding each block
51
       encodedmatrix =
52
          cellfun(@encodeDCTCoefficients,quantizedmatrix,'UniformOutput'
       ,false);
53
54
       \% Pointwise multiplication of quantizationMatrix and
55
          quantizedmatrix
       dequantizedmatrix = cellfun(@(x)
56
          x.*quantizationMatrix,quantizedmatrix,'UniformOutput',false);
57
       % Apply inverse DCT-2 to each block
58
       level_offed_matrix =
59
          cellfun(@idct2,dequantizedmatrix,'UniformOutput',false);
60
       \mbox{\ensuremath{\mbox{\%}}} Adding 128 to every single element
61
       decompressed_blocks = cellfun(@(x)
62
          x+128,level_offed_matrix,'UniformOutput',false);
63
       % Combine all blocks into a single image
64
       decompressed_image = cell2mat(decompressed_blocks);
65
66
       % Calculating Zero Percentage in the dequantizedmatrix
67
       dequantizedmatrix = cell2mat(dequantizedmatrix);
68
       zero_percentage =
69
          (nnz(dequantizedmatrix==0)/numel(dequantizedmatrix))*100;
70
71
       \% Calculating the maximum light intensity of the original image
          using antilog of one side of the image
       \max_{i=1}^{\infty} = 2^{(\log_2(rows))-1};
72
73
74
       % Calculating Mean Squared Error by iteratively using mean() over
75
          the two axes of a squared error matrix
       \% First, ensure that both original_image and decompressed_image
76
          have the same data type (e.g., double)
       original_image = double(original_image);
77
       decompressed_image = double(decompressed_image);
78
79
       % Now calculate the Mean Squared Error
80
```

```
mean_squared_error = mean(mean((original_image -
81
           decompressed_image).^2));
82
        % Calculating Peak Signal-to-Noise Ratio
83
        peak_signal_to_noise_ratio = 10 * log10((max_intensity^2) /
84
           mean_squared_error);
85
        % Display the calculated performance metrics
86
        temp = "When Quality Level = " + num2str(qualitylevel);
87
        disp(temp);
88
        pm1 = "Percentage of Zeros = " + num2str(zero_percentage)+"%";
89
        disp(pm1);
90
        pm2 = "Peak Signal-to-Noise Ratio(PSNR) = " +
91
           num2str(peak_signal_to_noise_ratio)+"dB";
        disp(pm2);
92
        disp(" ");
93
        % Displaying Images
94
        myfigure = figure();
95
        myfigure.WindowState = 'maximized';
96
97
        subplot(1, 3, 1);
98
        imshow(original_image, []);
99
        title('Original Image');
100
101
        subplot(1, 3, 2);
102
        imshow(cell2mat(quantizedmatrix), []);
103
        title_string = "Compressed Image at Quality Level = " +
104
           num2str(qualitylevel);
        title(title_string);
105
106
        subplot(1, 3, 3);
107
108
        imshow(decompressed_image, []);
        title_string = "Decompressed Image at Quality Level = " +
109
           num2str(qualitylevel);
        title(title_string);
110
   end
111
112
113
114
   function quantizationMatrix = createJPEGQuantizationMatrix(quality)
115
        \% Define the standard JPEG quantization matrix for quality level 50
116
        standardQuantizationMatrix = [
117
            16, 11, 10, 16, 24, 40, 51, 61;
118
            12, 12, 14, 19, 26, 58, 60, 55;
119
            14, 13, 16, 24, 40, 57, 69, 56;
120
```

```
14, 17, 22, 29, 51, 87, 80, 62;
121
            18, 22, 37, 56, 68, 109, 103, 77;
122
            24, 35, 55, 64, 81, 104, 113, 92;
123
            49, 64, 78, 87, 103, 121, 120, 101;
124
            72, 92, 95, 98, 112, 100, 103, 99
125
        ];
126
127
        % Quality should have values between 1 and 100
128
        if quality <= 0</pre>
129
            quality = 1;
130
        elseif quality > 100
131
            quality = 100;
132
        end
133
134
        \% Adjust Quantization Matrix based on the desired quality level
135
        if quality < 50
136
            scaleFactor = 50 / quality;
137
        else
138
            scaleFactor = 2 - (quality / 50);
139
        end
140
141
142
        quantizationMatrix = round(standardQuantizationMatrix *
           scaleFactor);
143
        % Ensure that no values are less than 1
144
        quantizationMatrix(quantizationMatrix < 1) = 1;</pre>
145
146
   end
147
   % Define a function to quantize the matrix with quantizationMatrix
148
149
   function quantizedmatrix = quantizeDCTCoefficients(matrix,
       quantizationMatrix)
150
        % Initialize the quantized matrix
        quantizedmatrix = round(matrix ./ quantizationMatrix);
151
   end
152
153
   % Define a function to encode [0,0] coefficient using differential PCM
154
       and encode AC coefficients with (run, level) pairs
   function encodedmatrix = encodeDCTCoefficients(matrix)
155
        % Initialize the encoded matrix
156
        encodedmatrix = zeros(1, 64);
157
158
        % Encode the DC coefficient
159
        encodedmatrix(1) = matrix(1, 1);
160
161
        % Encode the AC coefficients
162
```

```
index = 2;
163
         for i = 1:8
164
             for j = 1:8
165
                  if i == 1 && j == 1
166
                       continue;
167
                  end
168
                  if matrix(i, j) == 0
169
                       index = index + 1;
170
                  else
171
                       encodedmatrix(index) = matrix(i, j);
172
                       index = index + 1;
173
                  end
174
             end
175
         end
176
177
    end
```

Listing 2.2 — compress and observe image 1 ("camera512")

```
clc;
1
2
   clear;
   close all;
3
4
  % Loading the required image
5
   load("boat512.mat")
6
   image = boat512;
7
8
9
   [rows, cols, ~] = size(image);
10
11
   % Determine the number of 8x8 blocks needed along each dimension
12
   numBlocksRows = ceil(rows/8);
13
14
   numBlocksCols = ceil(cols/8);
15
   \% Make a copy of the original image before doing operations using it
16
   original_image = image;
17
18
   % Divide the image into 8x8 blocks
19
   image = mat2cell(image, 8 * ones(1, numBlocksRows), 8 * ones(1,
20
      numBlocksCols));
21
   % Subtract 128 from every single element
22
   image = cellfun(@(x) x - 128, image, 'UniformOutput', false);
23
   % Apply DCT-2 to each block
25
   image = cellfun(@dct2, image, 'UniformOutput', false);
26
27
```

```
% Define the quality levels for compression.
28
   qualityLevels = [60, 25, 10];
29
   numQualityLevels = length(qualityLevels);
30
31
   % Loop through each quality level.
32
   for level = 1:numQualityLevels
33
       % Calculate the quality factor for this level.
34
       qualitylevel = qualityLevels(level);
35
       % Defining Quantization Matrix
36
       quantizationMatrix = createJPEGQuantizationMatrix(qualitylevel);
37
38
       % Quantize each block
39
       quantizedmatrix = cellfun(@(x)
40
          quantizeDCTCoefficients(x,quantizationMatrix),image,
       'UniformOutput', false);
41
42
       % Combine quantizedmatrix into a single image
43
       quantizedmatrix = cell2mat(quantizedmatrix);
44
       quantizedmatrix_copy = quantizedmatrix;
45
46
       % Converting quantized matrix into a cell array
47
       quantizedmatrix =
48
          mat2cell(quantizedmatrix,8*ones(1,numBlocksRows),8*ones(1,
       numBlocksCols));
49
50
       % Encoding each block
51
       encodedmatrix =
52
          cellfun(@encodeDCTCoefficients,quantizedmatrix,'UniformOutput'
       ,false);
53
54
       % Pointwise multiplication of quantizationMatrix and
55
          quantizedmatrix
       dequantizedmatrix = cellfun(@(x)
56
          x.*quantizationMatrix,quantizedmatrix,'UniformOutput',false);
57
       % Apply inverse DCT-2 to each block
58
       level_offed_matrix =
59
          cellfun(@idct2,dequantizedmatrix,'UniformOutput',false);
60
       % Adding 128 to every single element
61
       decompressed_blocks = cellfun(@(x)
62
          x+128,level_offed_matrix,'UniformOutput',false);
63
       % Combine all blocks into a single image
64
       decompressed_image = cell2mat(decompressed_blocks);
65
```

```
66
        \% Calculating Zero Percentage in the dequantizedmatrix
67
        dequantizedmatrix = cell2mat(dequantizedmatrix);
68
        zero_percentage =
69
           (nnz(dequantizedmatrix==0)/numel(dequantizedmatrix))*100;
70
        % Calculating the maximum light intensity of the original image
71
           using antilog of one side of the image
        \max_{i=1}^{\infty} \frac{1}{2} (\log_{i}(2 \log n)) - 1;
72
73
74
        % Calculating Mean Squared Error by iteratively using mean() over
75
           the two axes of a squared error matrix
        % First, ensure that both original_image and decompressed_image
76
           have the same data type (e.g., double)
        original_image = double(original_image);
77
        decompressed_image = double(decompressed_image);
78
79
        % Now calculate the Mean Squared Error
80
        mean_squared_error = mean(mean((original_image -
81
           decompressed_image).^2));
82
        % Calculating Peak Signal-to-Noise Ratio
83
        peak_signal_to_noise_ratio = 10 * log10((max_intensity^2) /
84
           mean_squared_error);
85
        % Display the calculated performance metrics
86
        temp = "When Quality Level = " + num2str(qualitylevel);
87
        disp(temp);
88
89
        pm1 = "Percentage of Zeros = " + num2str(zero_percentage)+"%";
        disp(pm1);
90
91
        pm2 = "Peak Signal-to-Noise Ratio(PSNR) =
           num2str(peak_signal_to_noise_ratio)+"dB";
        disp(pm2);
92
       disp(" ");
93
        % Displaying Images
94
        myfigure = figure();
95
        myfigure.WindowState = 'maximized';
96
97
        subplot(1, 3, 1);
98
        imshow(original_image, []);
99
        title('Original Image');
100
101
        subplot(1, 3, 2);
102
        imshow(cell2mat(quantizedmatrix), []);
103
```

```
title_string = "Compressed Image at Quality Level = " +
104
           num2str(qualitylevel);
        title(title_string);
105
106
        subplot(1, 3, 3);
107
        imshow(decompressed_image, []);
108
        title_string = "Decompressed Image at Quality Level = " +
109
           num2str(qualitylevel);
        title(title_string);
110
   end
111
112
113
114
115
   function quantizationMatrix = createJPEGQuantizationMatrix(quality)
        \% Define the standard JPEG quantization matrix for quality level 50
116
        standardQuantizationMatrix = [
117
            16, 11, 10, 16, 24, 40, 51, 61;
118
            12, 12, 14, 19, 26, 58, 60, 55;
119
            14, 13, 16, 24, 40, 57, 69, 56;
120
            14, 17, 22, 29, 51, 87, 80, 62;
121
            18, 22, 37, 56, 68, 109, 103, 77;
122
            24, 35, 55, 64, 81, 104, 113, 92;
123
            49, 64, 78, 87, 103, 121, 120, 101;
124
            72, 92, 95, 98, 112, 100, 103, 99
125
        ];
126
127
        % Quality should have values between 1 and 100
128
        if quality <= 0</pre>
129
            quality = 1;
130
        elseif quality > 100
131
            quality = 100;
132
133
        end
134
        % Adjust Quantization Matrix based on the desired quality level
135
        if quality < 50
136
            scaleFactor = 50 / quality;
137
        else
138
            scaleFactor = 2 - (quality / 50);
139
        end
140
141
        quantizationMatrix = round(standardQuantizationMatrix *
142
           scaleFactor);
143
        % Ensure that no values are less than 1
144
        quantizationMatrix(quantizationMatrix < 1) = 1;</pre>
145
```

```
end
146
147
   \% Define a function to quantize the matrix with quantizationMatrix
148
   function quantizedmatrix = quantizeDCTCoefficients(matrix,
149
       quantizationMatrix)
        % Initialize the quantized matrix
150
        quantizedmatrix = round(matrix ./ quantizationMatrix);
151
   end
152
153
   % Define a function to encode [0,0] coefficient using differential PCM
154
       and encode AC coefficients with (run, level) pairs
   function encodedmatrix = encodeDCTCoefficients(matrix)
155
        % Initialize the encoded matrix
156
        encodedmatrix = zeros(1, 64);
157
158
        % Encode the DC coefficient
159
        encodedmatrix(1) = matrix(1, 1);
160
161
        % Encode the AC coefficients
162
        index = 2;
163
        for i = 1:8
164
            for j = 1:8
165
                 if i == 1 && j == 1
166
                     continue;
167
168
                 if matrix(i, j) == 0
169
                     index = index + 1;
170
                 else
171
                     encodedmatrix(index) = matrix(i, j);
172
173
                     index = index + 1;
                 end
174
175
            end
        end
176
177
   end
```

Listing 2.3 — compress and observe image 2 ("boat512")

```
clc;
clear;
close all;

% Loading the required image
load("peppers512.mat")
image = peppers512;

[rows, cols, ~] = size(image);
```

```
10
   % Determine the number of 8x8 blocks needed along each dimension
11
   numBlocksRows = ceil(rows/8);
12
   numBlocksCols = ceil(cols/8);
13
14
   \% Make a copy of the original image before doing operations using it
15
   original_image = image;
16
17
   % Divide the image into 8x8 blocks
18
   image = mat2cell(image, 8 * ones(1, numBlocksRows), 8 * ones(1,
19
      numBlocksCols));
20
   % Subtract 128 from every single element
21
   image = cellfun(@(x) x - 128, image, 'UniformOutput', false);
22
23
   % Apply DCT-2 to each block
24
   image = cellfun(@dct2, image, 'UniformOutput', false);
25
26
   \% Define the quality levels for compression.
27
   qualityLevels = [60, 25, 10];
28
   numQualityLevels = length(qualityLevels);
29
30
   % Loop through each quality level.
31
   for level = 1:numQualityLevels
32
       % Calculate the quality factor for this level.
33
       qualitylevel = qualityLevels(level);
34
       % Defining Quantization Matrix
35
       quantizationMatrix = createJPEGQuantizationMatrix(qualitylevel);
36
37
38
       % Quantize each block
       quantizedmatrix = cellfun(@(x)
39
          quantizeDCTCoefficients(x,quantizationMatrix), image,
       'UniformOutput', false);
40
41
       % Combine quantizedmatrix into a single image
42
       quantizedmatrix = cell2mat(quantizedmatrix);
43
       quantizedmatrix_copy = quantizedmatrix;
44
45
       \% Converting quantized matrix into a cell array
46
       quantizedmatrix =
47
          mat2cell(quantizedmatrix, 8*ones(1, numBlocksRows), 8*ones(1,
       numBlocksCols));
48
49
       % Encoding each block
50
```

```
encodedmatrix =
51
          \verb|cellfun(@encodeDCTCoefficients, quantized matrix, 'UniformOutput')| \\
       ,false);
52
53
       % Pointwise multiplication of quantizationMatrix and
54
          quantizedmatrix
       dequantizedmatrix = cellfun(@(x)
55
          x.*quantizationMatrix,quantizedmatrix,'UniformOutput',false);
56
       % Apply inverse DCT-2 to each block
57
       level_offed_matrix =
58
          cellfun(@idct2,dequantizedmatrix,'UniformOutput',false);
59
       % Adding 128 to every single element
60
       decompressed_blocks = cellfun(@(x)
61
          x+128,level_offed_matrix,'UniformOutput',false);
62
       % Combine all blocks into a single image
63
       decompressed_image = cell2mat(decompressed_blocks);
64
65
       % Calculating Zero Percentage in the dequantizedmatrix
66
       dequantizedmatrix = cell2mat(dequantizedmatrix);
67
       zero_percentage =
68
          (nnz(dequantizedmatrix==0)/numel(dequantizedmatrix))*100;
69
       \% Calculating the maximum light intensity of the original image
70
          using antilog of one side of the image
       \max_{i=1}^{\infty} 1 = 2^{(\log_{2}(rows))-1};
71
72
73
       % Calculating Mean Squared Error by iteratively using mean() over
74
          the two axes of a squared error matrix
       % First, ensure that both original_image and decompressed_image
75
          have the same data type (e.g., double)
       original_image = double(original_image);
76
       decompressed_image = double(decompressed_image);
77
78
       % Now calculate the Mean Squared Error
79
       mean_squared_error = mean(mean((original_image -
80
          decompressed_image).^2));
81
       % Calculating Peak Signal-to-Noise Ratio
82
       peak_signal_to_noise_ratio = 10 * log10((max_intensity^2) /
83
          mean_squared_error);
84
```

```
% Display the calculated performance metrics
85
        temp = "When Quality Level = " + num2str(qualitylevel);
86
        disp(temp);
87
        pm1 = "Percentage of Zeros = " + num2str(zero_percentage)+"%";
88
        disp(pm1);
89
        pm2 = "Peak Signal-to-Noise Ratio(PSNR) = " +
90
           num2str(peak_signal_to_noise_ratio)+"dB";
        disp(pm2);
91
        disp(" ");
92
        % Displaying Images
93
        myfigure = figure();
94
        myfigure.WindowState = 'maximized';
95
96
        subplot(1, 3, 1);
97
        imshow(original_image, []);
98
        title('Original Image');
99
100
101
        subplot(1, 3, 2);
        imshow(cell2mat(quantizedmatrix), []);
102
        title_string = "Compressed Image at Quality Level = " +
103
           num2str(qualitylevel);
104
        title(title_string);
105
106
        subplot(1, 3, 3);
        imshow(decompressed_image, []);
107
        title_string = "Decompressed Image at Quality Level = " +
108
           num2str(qualitylevel);
109
        title(title_string);
   end
110
111
112
113
   function quantizationMatrix = createJPEGQuantizationMatrix(quality)
114
        \% Define the standard JPEG quantization matrix for quality level 50
115
        standardQuantizationMatrix = [
116
            16, 11, 10, 16, 24, 40, 51, 61;
117
            12, 12, 14, 19, 26, 58, 60, 55;
118
            14, 13, 16, 24, 40, 57, 69, 56;
119
            14, 17, 22, 29, 51, 87, 80, 62;
120
            18, 22, 37, 56, 68, 109, 103, 77;
121
            24, 35, 55, 64, 81, 104, 113, 92;
122
            49, 64, 78, 87, 103, 121, 120, 101;
123
            72, 92, 95, 98, 112, 100, 103, 99
124
        ];
125
126
```

```
% Quality should have values between 1 and 100
127
        if quality <= 0</pre>
128
            quality = 1;
129
        elseif quality > 100
130
            quality = 100;
131
        end
132
133
        \% Adjust Quantization Matrix based on the desired quality level
134
        if quality < 50</pre>
135
            scaleFactor = 50 / quality;
136
        else
137
            scaleFactor = 2 - (quality / 50);
138
        end
139
140
        quantizationMatrix = round(standardQuantizationMatrix *
141
           scaleFactor);
142
        % Ensure that no values are less than 1
143
        quantizationMatrix(quantizationMatrix < 1) = 1;</pre>
144
   end
145
146
147
   % Define a function to quantize the matrix with quantizationMatrix
   function quantizedmatrix = quantizeDCTCoefficients(matrix,
148
       quantizationMatrix)
        % Initialize the quantized matrix
149
        quantizedmatrix = round(matrix ./ quantizationMatrix);
150
151
152
   % Define a function to encode [0,0] coefficient using differential PCM
153
       and encode AC coefficients with (run, level) pairs
   function encodedmatrix = encodeDCTCoefficients(matrix)
154
155
        % Initialize the encoded matrix
        encodedmatrix = zeros(1, 64);
156
157
        % Encode the DC coefficient
158
        encodedmatrix(1) = matrix(1, 1);
159
160
        % Encode the AC coefficients
161
        index = 2;
162
        for i = 1:8
163
            for j = 1:8
164
                 if i == 1 && j == 1
165
                     continue;
166
                 end
167
                 if matrix(i, j) == 0
168
```

```
index = index + 1;
169
                  else
170
                       encodedmatrix(index) = matrix(i, j);
171
                       index = index + 1;
172
                  end
173
              end
174
         end
175
    end
176
```

Listing 2.4 — compress and observe image 3 ("peppers512")

```
clc;
1
   clear;
   close all;
3
4
  % Loading the required image
5
   image = imread("dog.jpg");
6
   [rows, cols, ~] = size(image);
7
8
9
   % Determine the number of 8x8 blocks needed along each dimension
   numBlocksRows = ceil(rows/8);
10
   numBlocksCols = ceil(cols/8);
11
12
13
   \% Make a copy of the original image before doing operations using it
   original_image = image;
14
15
   % Divide the image into 8x8 blocks
16
   image = mat2cell(image, 8 * ones(1, numBlocksRows), 8 * ones(1,
17
      numBlocksCols));
18
   % Subtract 128 from every single element
19
   image = cellfun(@(x) x - 128, image, 'UniformOutput', false);
20
21
   % Apply DCT-2 to each block
22
   image = cellfun(@dct2, image, 'UniformOutput', false);
23
24
   \% Define the quality levels for compression.
25
   qualityLevels = [60, 25, 10];
26
   numQualityLevels = length(qualityLevels);
27
28
   % Loop through each quality level.
29
   for level = 1:numQualityLevels
30
       \% Calculate the quality factor for this level.
31
       qualitylevel = qualityLevels(level);
32
       % Defining Quantization Matrix
33
       quantizationMatrix = createJPEGQuantizationMatrix(qualitylevel);
34
```

```
35
       % Quantize each block
36
       quantizedmatrix = cellfun(@(x)
37
          quantizeDCTCoefficients(x,quantizationMatrix), image,
       'UniformOutput', false);
38
39
       % Combine quantizedmatrix into a single image
40
       quantizedmatrix = cell2mat(quantizedmatrix);
41
       quantizedmatrix_copy = quantizedmatrix;
42
43
       % Converting quantized matrix into a cell array
44
       quantizedmatrix =
45
          mat2cell(quantizedmatrix, 8*ones(1, numBlocksRows), 8*ones(1,
       numBlocksCols));
46
47
       % Encoding each block
48
       encodedmatrix =
49
          cellfun(@encodeDCTCoefficients,quantizedmatrix,'UniformOutput'
       ,false);
50
51
       % Pointwise multiplication of quantizationMatrix and
52
          quantizedmatrix
       dequantizedmatrix = cellfun(@(x)
53
          x.*quantizationMatrix,quantizedmatrix,'UniformOutput',false);
54
       % Apply inverse DCT-2 to each block
55
       level_offed_matrix =
56
          cellfun(@idct2,dequantizedmatrix,'UniformOutput',false);
57
       % Adding 128 to every single element
58
       decompressed_blocks = cellfun(@(x)
59
          x+128,level_offed_matrix,'UniformOutput',false);
60
       % Combine all blocks into a single image
61
       decompressed_image = cell2mat(decompressed_blocks);
62
63
       % Calculating Zero Percentage in the dequantizedmatrix
64
       dequantizedmatrix = cell2mat(dequantizedmatrix);
65
       zero_percentage =
66
           (nnz(dequantizedmatrix==0)/numel(dequantizedmatrix))*100;
67
       % Calculating the maximum light intensity of the original image
68
          using antilog of one side of the image
       \max_{i=1}^{\infty} \frac{1}{2} (\log_{i}(2 \cos i)) - 1;
69
70
```

```
71
       % Calculating Mean Squared Error by iteratively using mean() over
72
           the two axes of a squared error matrix
       % First, ensure that both original_image and decompressed_image
73
           have the same data type (e.g., double)
       original_image = double(original_image);
74
       decompressed_image = double(decompressed_image);
75
76
       % Now calculate the Mean Squared Error
77
       mean_squared_error = mean(mean((original_image -
78
           decompressed_image).^2));
79
       % Calculating Peak Signal-to-Noise Ratio
80
       peak_signal_to_noise_ratio = 10 * log10((max_intensity^2) /
81
           mean_squared_error);
82
       % Display the calculated performance metrics
83
       temp = "When Quality Level = " + num2str(qualitylevel);
84
       disp(temp);
85
       pm1 = "Percentage of Zeros = " + num2str(zero_percentage)+"%";
86
       disp(pm1);
87
       pm2 = "Peak Signal-to-Noise Ratio(PSNR) = " +
88
           num2str(peak_signal_to_noise_ratio)+"dB";
       disp(pm2);
89
       disp(" ");
90
       % Displaying Images
91
       myfigure = figure();
92
       myfigure.WindowState = 'maximized';
93
94
95
       subplot(1, 3, 1);
       imshow(original_image, []);
96
97
       title('Original Image');
98
       subplot(1, 3, 2);
99
       imshow(cell2mat(quantizedmatrix), []);
100
       title_string = "Compressed Image at Quality Level = " +
101
           num2str(qualitylevel);
       title(title_string);
102
103
       subplot(1, 3, 3);
104
       imshow(decompressed_image, []);
105
       title_string = "Decompressed Image at Quality Level = " +
106
           num2str(qualitylevel);
       title(title_string);
107
108
   end
```

```
109
110
111
   function quantizationMatrix = createJPEGQuantizationMatrix(quality)
112
        \% Define the standard JPEG quantization matrix for quality level 50
113
        standardQuantizationMatrix = [
114
            16, 11, 10, 16, 24, 40, 51, 61;
115
            12, 12, 14, 19, 26, 58, 60, 55;
116
            14, 13, 16, 24, 40, 57, 69, 56;
117
            14, 17, 22, 29, 51, 87, 80, 62;
118
            18, 22, 37, 56, 68, 109, 103, 77;
119
            24, 35, 55, 64, 81, 104, 113, 92;
120
            49, 64, 78, 87, 103, 121, 120, 101;
121
            72, 92, 95, 98, 112, 100, 103, 99
122
        ];
123
124
        % Quality should have values between 1 and 100
125
        if quality <= 0</pre>
126
            quality = 1;
127
        elseif quality > 100
128
            quality = 100;
129
130
        end
131
        \% Adjust Quantization Matrix based on the desired quality level
132
        if quality < 50
133
            scaleFactor = 50 / quality;
134
        else
135
            scaleFactor = 2 - (quality / 50);
136
        end
137
138
        quantizationMatrix = round(standardQuantizationMatrix *
139
           scaleFactor);
140
        % Ensure that no values are less than 1
141
        quantizationMatrix(quantizationMatrix < 1) = 1;</pre>
142
   end
143
144
   % Define a function to quantize the matrix with quantizationMatrix
145
   function quantizedmatrix = quantizeDCTCoefficients(matrix,
146
       quantizationMatrix)
        % Initialize the quantized matrix
147
        quantizedmatrix = round(matrix ./ quantizationMatrix);
148
   end
149
150
```

```
% Define a function to encode [0,0] coefficient using differential PCM
151
       and encode AC coefficients with (run, level) pairs
   function encodedmatrix = encodeDCTCoefficients(matrix)
152
        % Initialize the encoded matrix
153
        encodedmatrix = zeros(1, 64);
154
155
        % Encode the DC coefficient
156
        encodedmatrix(1) = matrix(1, 1);
157
158
        % Encode the AC coefficients
159
        index = 2;
160
        for i = 1:8
161
            for j = 1:8
162
163
                 if i == 1 && j == 1
                     continue;
164
165
                 if matrix(i, j) == 0
166
                     index = index + 1;
167
                 else
168
                     encodedmatrix(index) = matrix(i, j);
169
                     index = index + 1;
170
171
                 end
            end
172
173
        end
   end
174
```

Listing 2.5 — compress and observe image 4("dog")
