

Comp Eng 4TN4

Project 1: Image Enhancement

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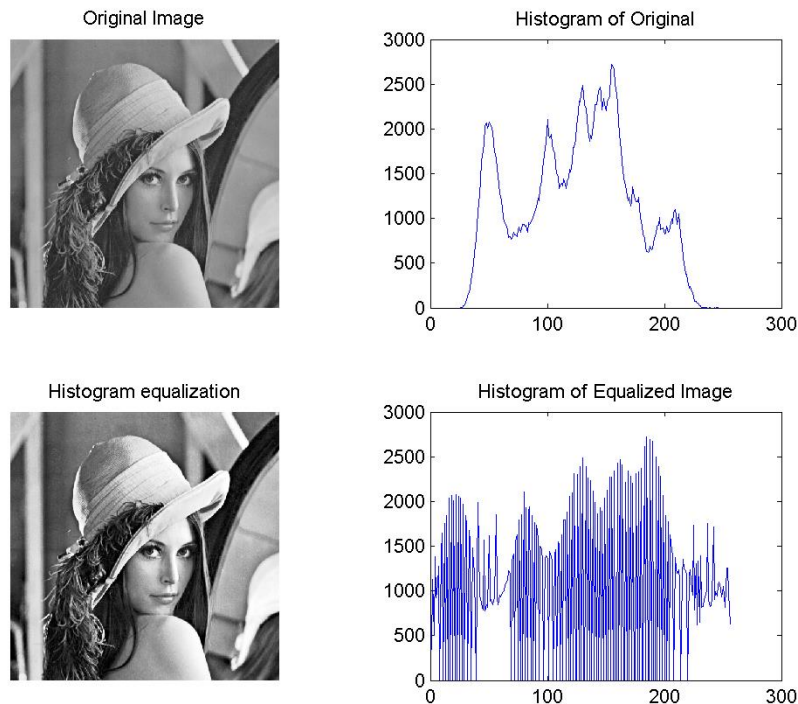
Histogram Equalization

Histogram Equalization (HE) is an image enhancement technique that helps evenly distribute image intensities to enhance contrast. Histogram equalization uses the cumulative distribution function (CDF), as shown below.

$$p_x(i) = p(x = i) = \frac{n_i}{n}, \quad 0 \leq i < L$$
$$cdf_x(i) = \sum_{j=0}^i p_x(j)$$

The transform that is used in histogram equalization is a single-valued and monotonically increasing. This forces the mapping of the lowest gray-level to zero and the highest gray-level value to the maximum gray-level, which can be determined by the number of bits used to represent pixel intensity. This results in the gray-levels to be distributed to span across missing gray-level values of gray-levels or less occurring gray-level values. Therefore, the details that may have been previously difficult to detect due to small changes of intensity values, they may become easier to discern as the gray-levels are spread to a larger range. In this way the enhanced image will display a higher contrast.

Test 1:



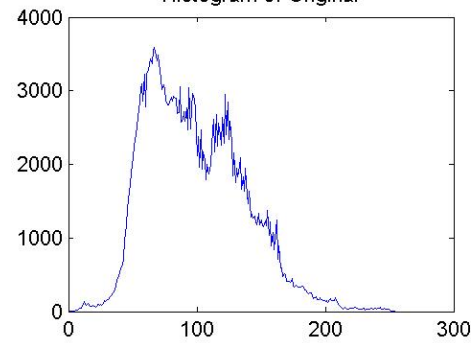
The following test cases show examples of histogram equalization that produce relatively desirable results. The contrast of the images is increased and more details are noticeable.

Test 2:

Original Image



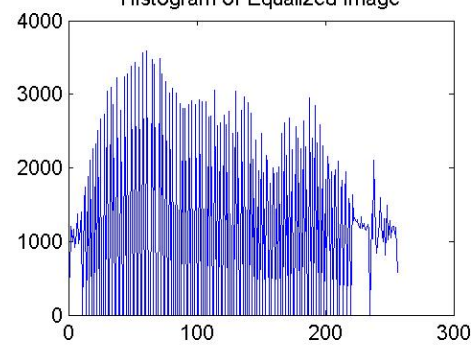
Histogram of Original



Histogram equalization



Histogram of Equalized Image



Test 3:

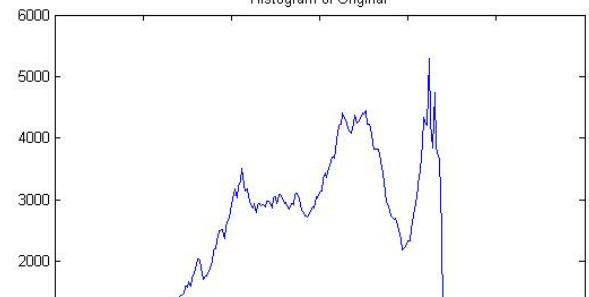
Original Image



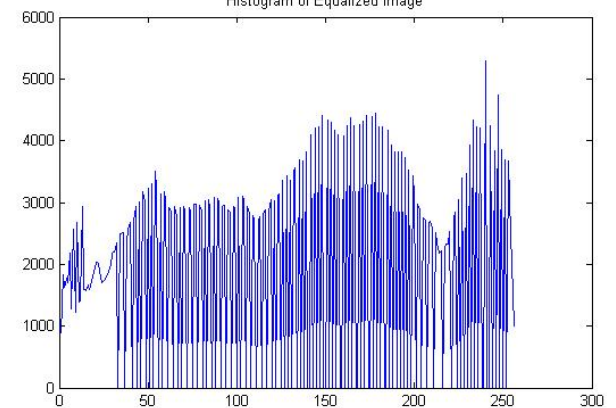
Histogram equalization



Histogram of Original

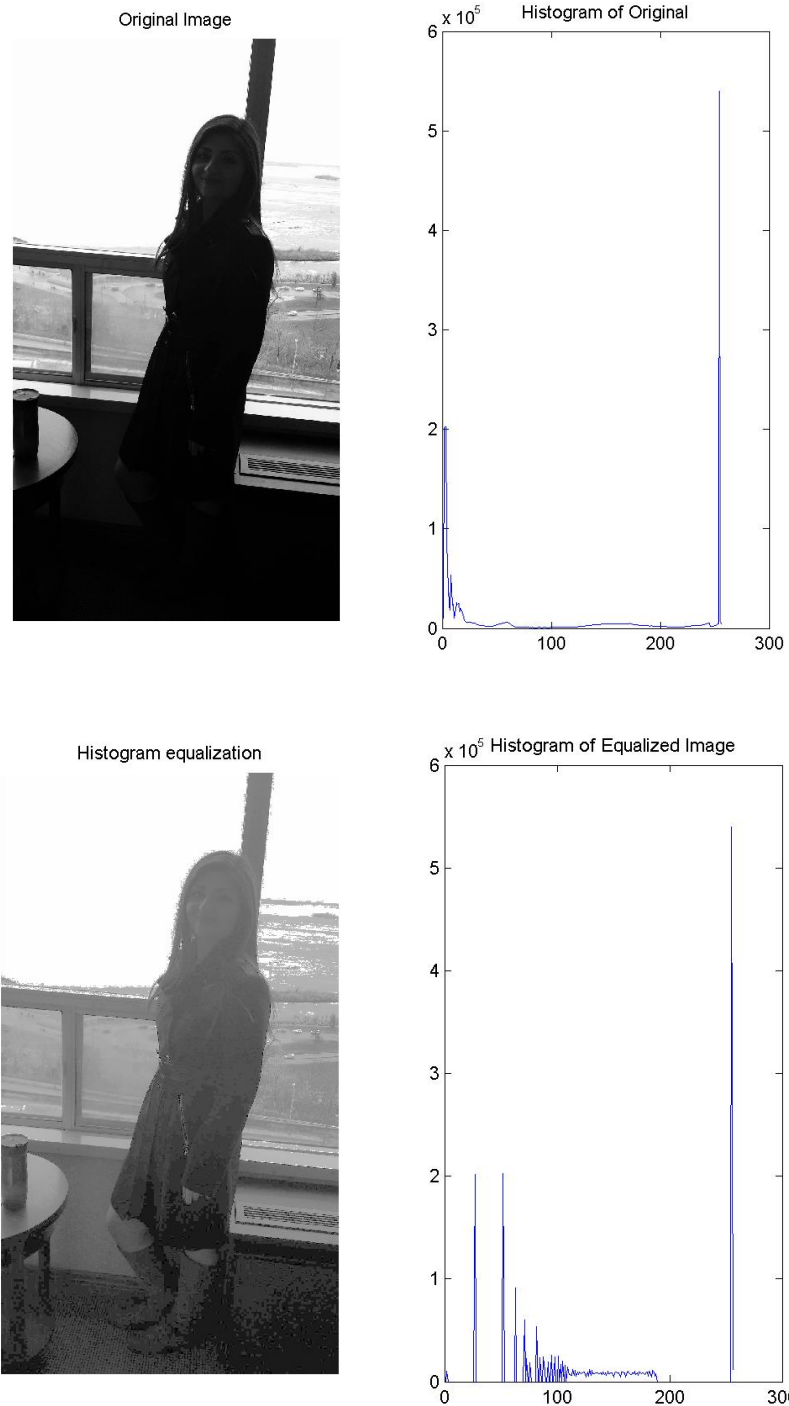


Histogram of Equalized Image



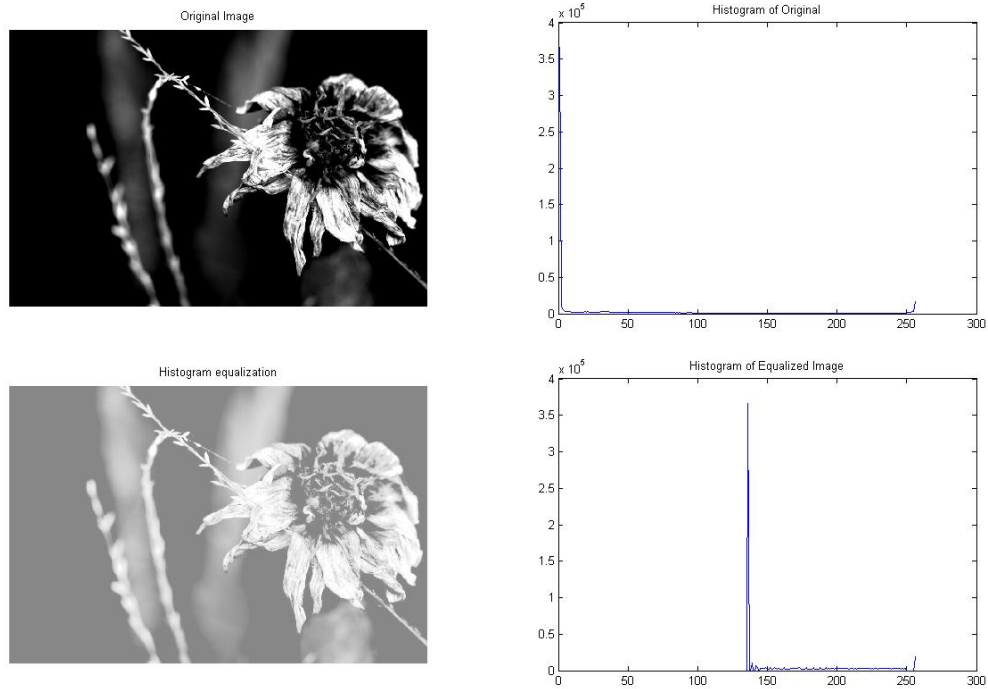
The following test case shows an example of undesirable histogram equalization output. This image already contains high contrast, and when histogram equalization is applied, the levels are spread out to more gray levels, resulting in a distorted image, where neither the foreground nor the background is enhanced.

Test 4:

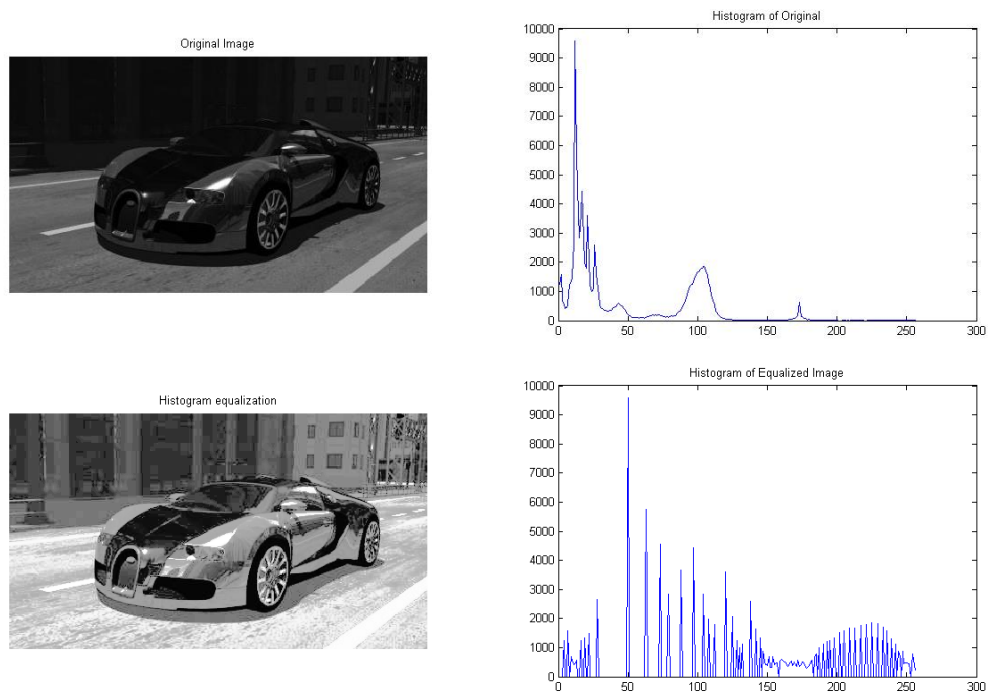


This high contrast image becomes equalized to have a more washed-out appearance. Here, the contrast is lost.

Test 5:



Test 6:



The image of the car was able to make certain ranges more visible; however, the image appears to be pixelated and generally becomes too bright, losing its original tone.

Optimal Contrast Tone Mapping

Optimal Contrast Tone Mapping (OCTM) is a method that has the same goal of contrast stretching as in histogram equalization. However, OCTM also applies constraints to ensure that the tone of the image is not lost in the process. OCTM utilizes linear programming optimisation in which an objective function, such as contrast, is maximized subject to constraints. These constraints may include the limited number of grayscale values that the pixels can take due to the fact that the gray-levels are determined by the number of bits used to represent pixel intensity. Another constraint is limiting the tone distortion so as to not have many input gray-levels being mapped onto the same output gray-level.

There are several conditions that are placed on the output of the OCTM images. Firstly, an integer-to-integer mapping is required, since the same number of gray-levels is to be maintained, and is discrete.

$$T : \{0, 1, \dots, 255\} \rightarrow \{0, 1, \dots, 255\}$$

T is the transfer function, and the values represent the intensity values. Secondly, the condition for monotonicity must also be met as described by the following equations.

$$\begin{aligned} T(j) &\geq T(i) \quad \text{if } j > i \\ T(i) &= \sum_{0 \leq j \leq i} s_j, \quad 0 \leq i < 256 \\ 0 &\leq s_j < 256 \\ \sum_{0 \leq j < 256} s_j &< 256 \end{aligned}$$

s_j is the increment in output intensity versus a unit step up in input level j (i.e. the contrast level at j). The expected contrast level of the image is given by the following equation.

$$C(s) = \sum_{0 \leq j < N_x} p_j s_j$$

p_j is the probability of the gray-level j . This is the objective function that must be maximized through linear programming methods. The following represent the constraints of the optimization problem.

$$\begin{aligned} &\max_s \sum_{0 \leq j < L} p_j s_j \\ \text{subject to (a)} &\sum_{0 \leq j < L} s_j < L; \\ &\text{(b) } s_j \geq 0, \quad 0 \leq j < L; \\ &\text{(c) } \sum_{j \leq i < j+d} s_i \geq 1, \quad 0 \leq j < L - d. \end{aligned}$$

- a) Confines the output intensity level to the available dynamic range
- b) Ensures the monotonicity condition of the transfer function
- c) Specifies the maximum tone distortion allowed by setting d at the upper limit

The contrast-tone optimization problem can therefore be stated by the following statement and the transfer function can be calculated using the equation below.

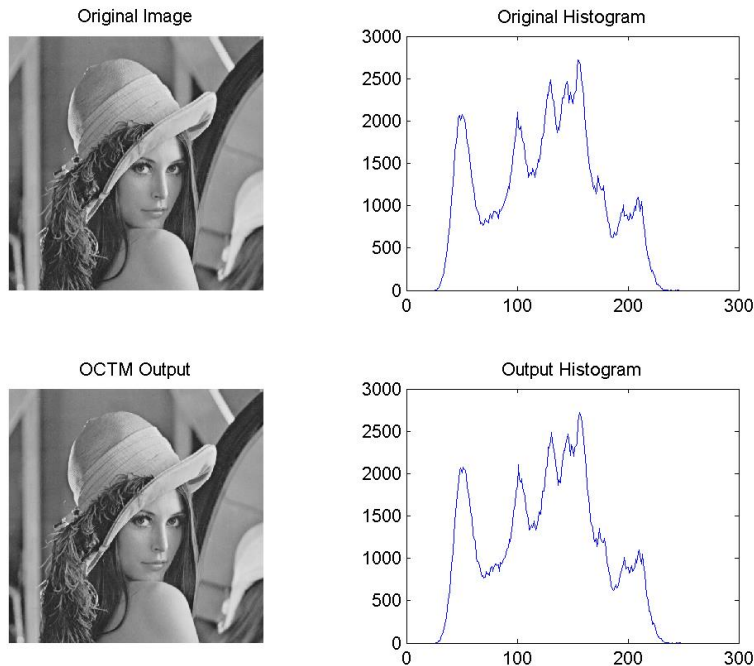
$$\begin{aligned} & \max_{\mathbf{s}} \sum_{0 \leq j < L} p_j s_j \\ & \text{subject to } \sum_{0 \leq j < L} s_j < L; \\ & s_j \geq 1/d, \quad 0 \leq j < L. \end{aligned}$$

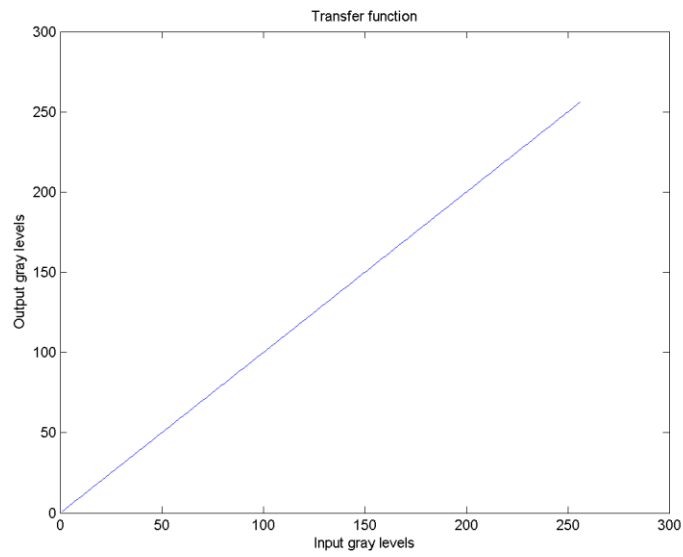
$$T(i) = \left\lfloor \sum_{0 \leq j \leq i} s_j + 0.5 \right\rfloor, \quad 0 \leq i < L$$

This problem can be solved using the optimization toolbox linear programming function on Matlab; `linprog(f,A,b,lower_bound,upper_bound)`, where f is the objective function, A and b are the inequality constraints imposed on the objective function, and the upper bound and lower bound of the problem is specified. The function will then solve for S_j .

The following 3 test cases apply the constraints stated above to try to produce enhanced images. There are 3 examples of different values of distortion that were tested. When $d = 1$, the image produced looks the same as the original image. As d increases, so does the contrast of the image, however, some distortion in the tone is noticeable.

Test 1: $d = 1$



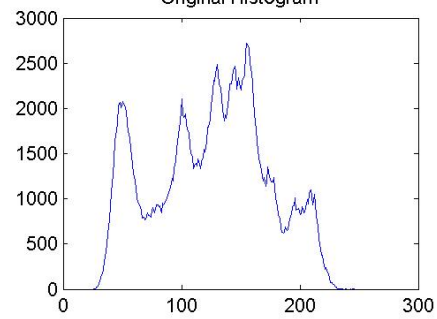


Test 2: $d = 1.1$

Original Image



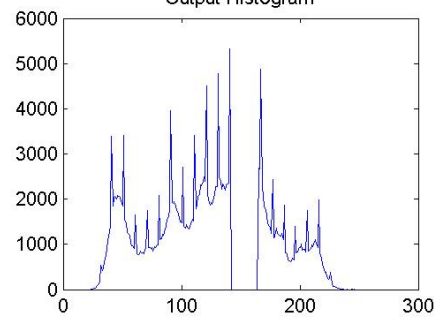
Original Histogram

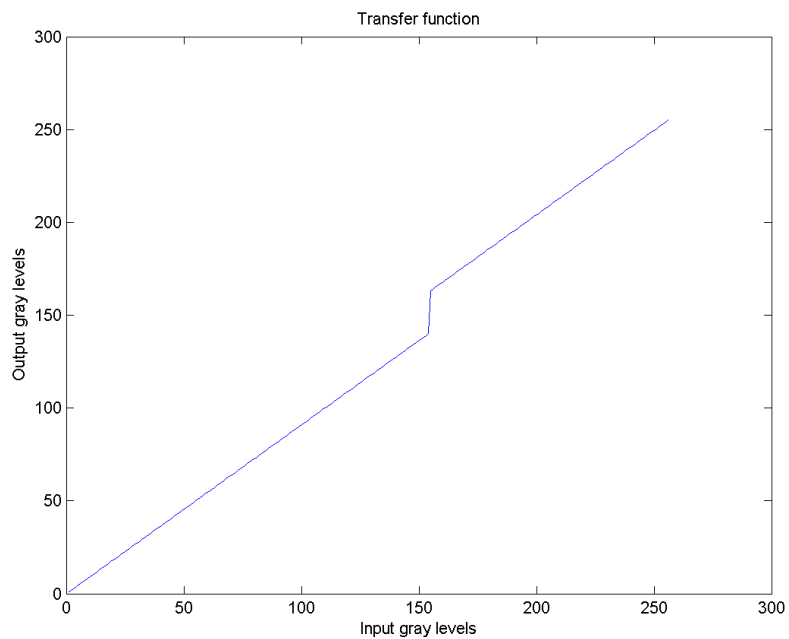


OCTM Output



Output Histogram



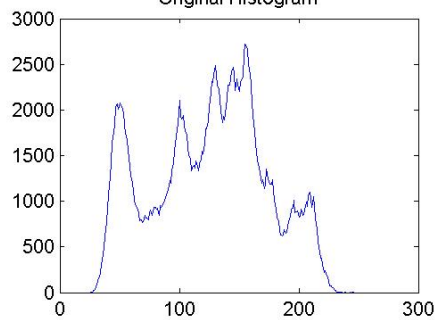


Test 3: $d = 1.5$

Original Image



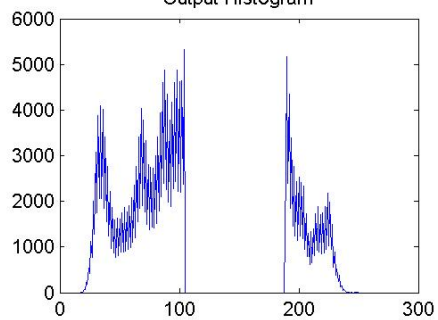
Original Histogram

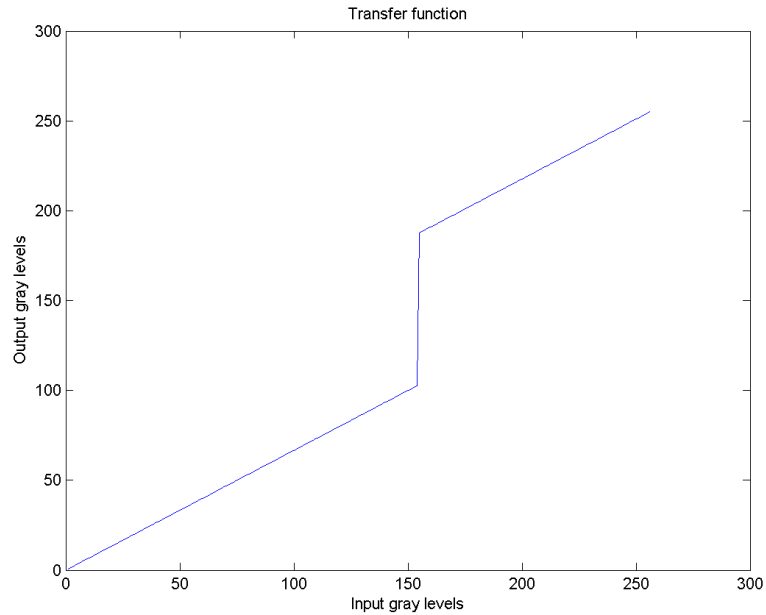


OCTM Output



Output Histogram





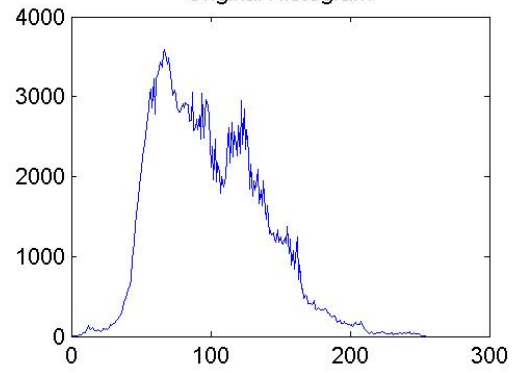
The following 4 test cases show the results of a different image with various values of distortion specified in the lower bound constraint. When the value of $d < 1$, the transfer function shows that the output level goes beyond the maximum value, therefore there is clipping of the image and it appears to be brighter than intended. When the value of $d > 1$, the transfer function maps correctly but some distortion may be visible at high values of d .

Test 4: $d = 0.8$

Original Image



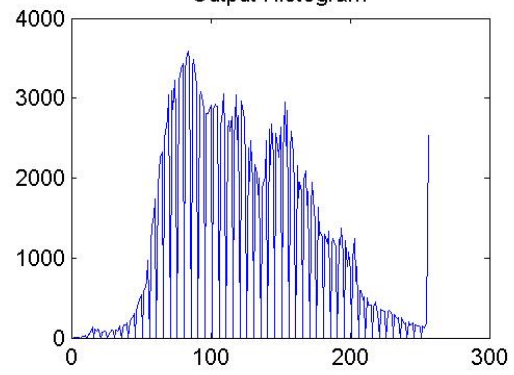
Original Histogram



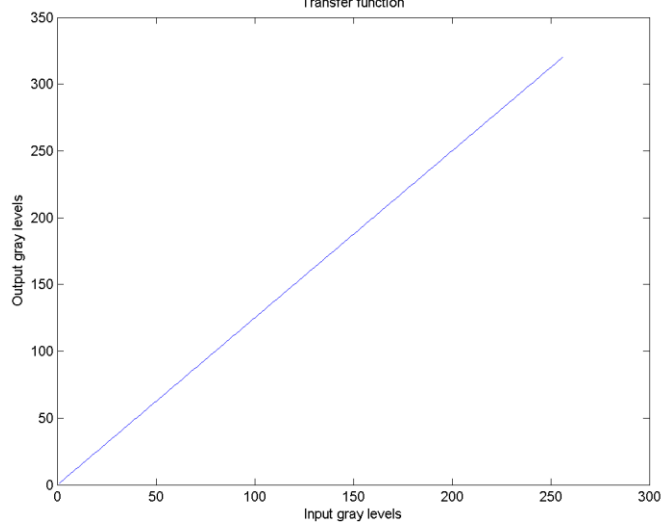
OCTM Output



Output Histogram



Transfer function

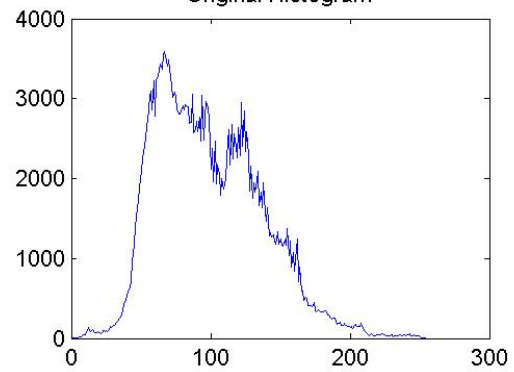


Test5: $d = 0.9$

Original Image



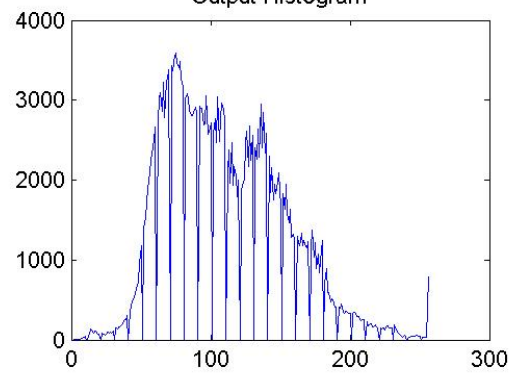
Original Histogram



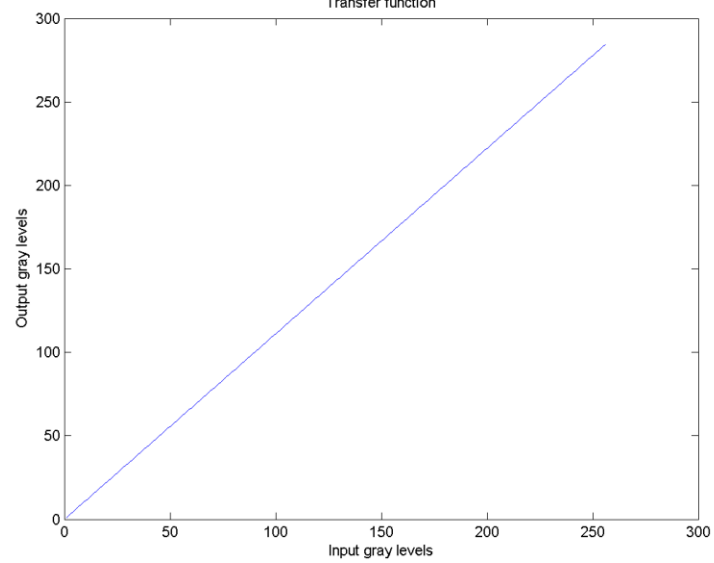
OCTM Output



Output Histogram



Transfer function

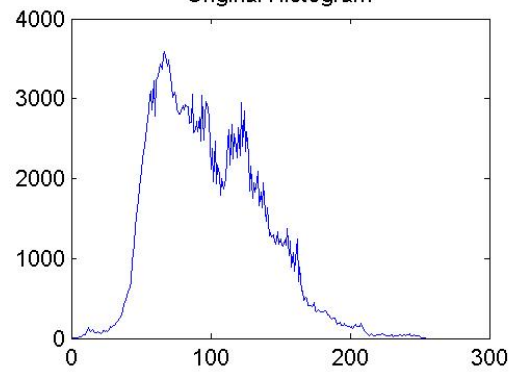


Test 6: $d = 1.1$

Original Image



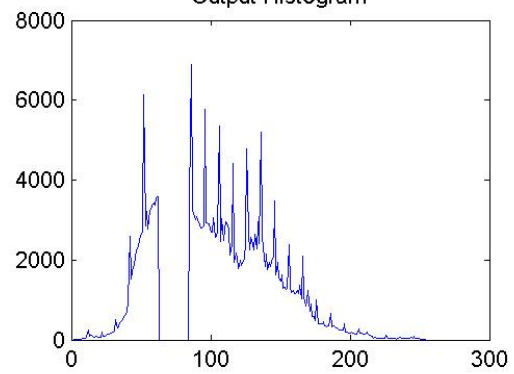
Original Histogram



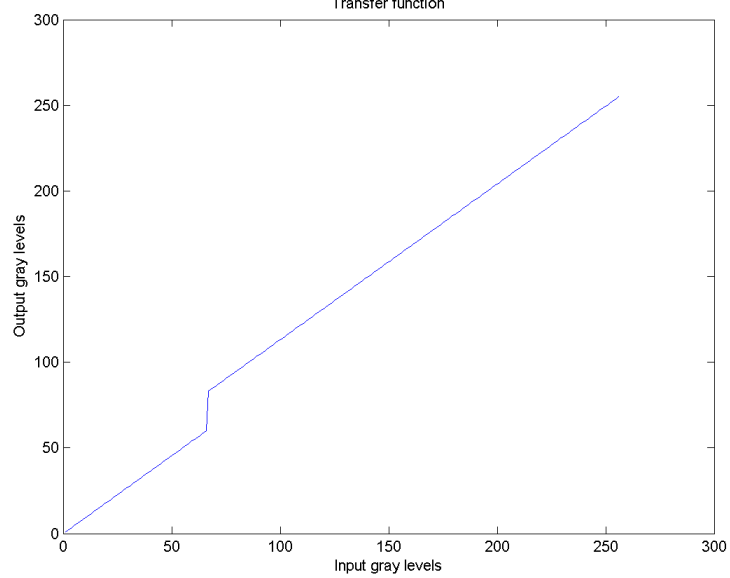
OCTM Output



Output Histogram



Transfer function

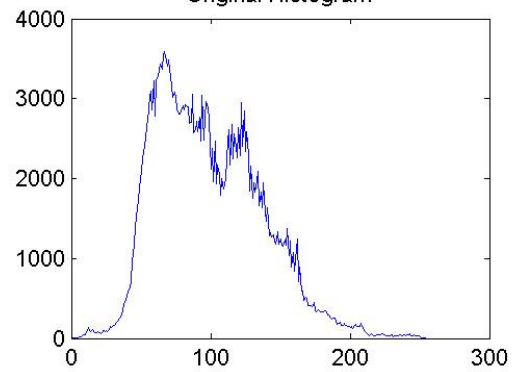


Test 7: $d = 1.2$

Original Image



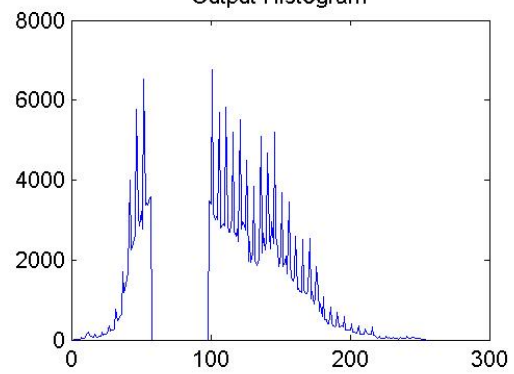
Original Histogram



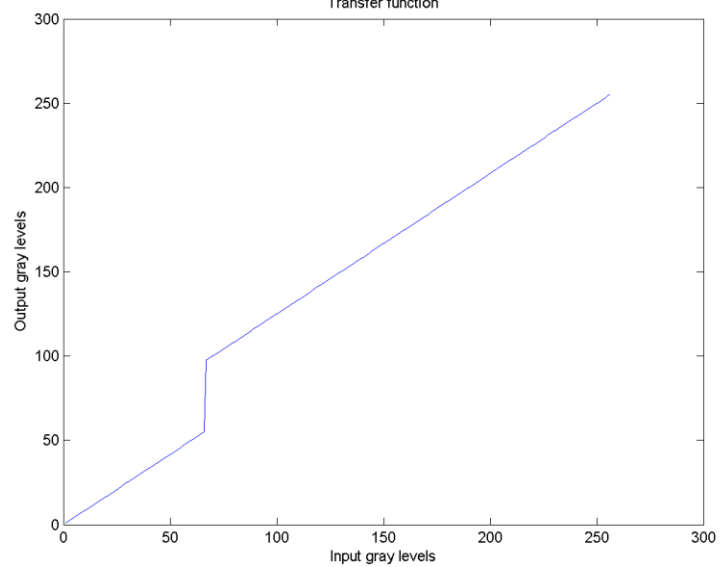
OCTM Output



Output Histogram



Transfer function



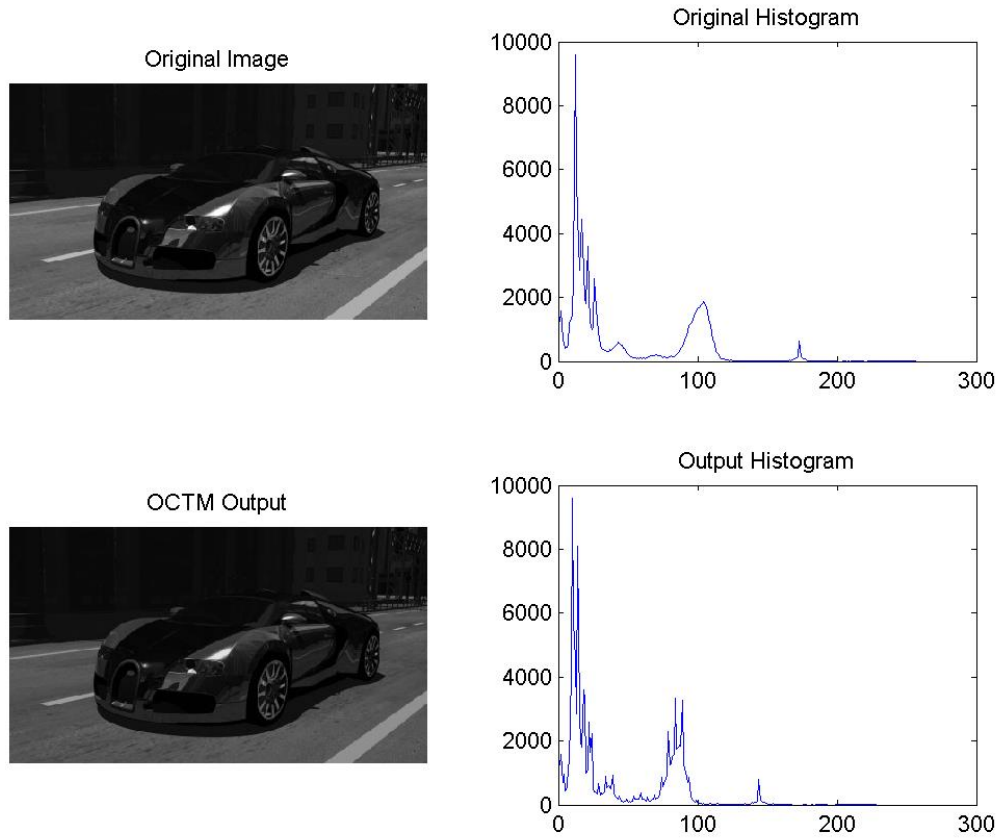
Additional constraints – Average intensity

Some additional constraints may be applied to OCTM in order to improve the results of the images. In this case, the average intensity of the image was taken into consideration. The restriction can be added by limiting the intensity of the output image to be within a certain range of the original input image. This is shown by the following equations:

$$\left| \frac{L}{t} \sum_{0 \leq i < L} p_i \sum_{0 \leq j \leq i} s_j - \sum_{0 \leq i < L} p_i i \right| \leq \Delta_\mu$$
$$\sum_{0 \leq i < L} \left| (L-1)^{-1} \sum_{0 \leq j \leq i} s_j - [i(L-1)^{-1}]^\gamma \right| \leq \Delta$$

In the Matlab implementation, the value of the range the intensity can take is represented by the variable *diff*.

Test 1: $d = 3$, $\text{diff} = 0$

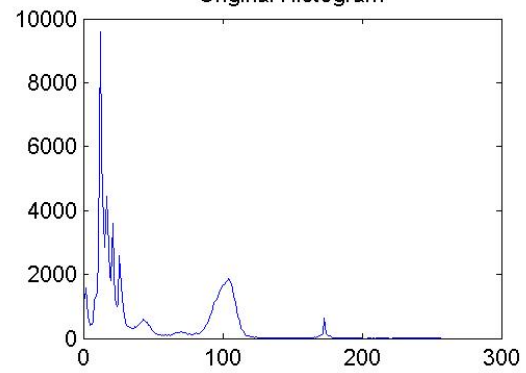


Test 2: $d = 3$, $\text{diff} = 1$

Original Image



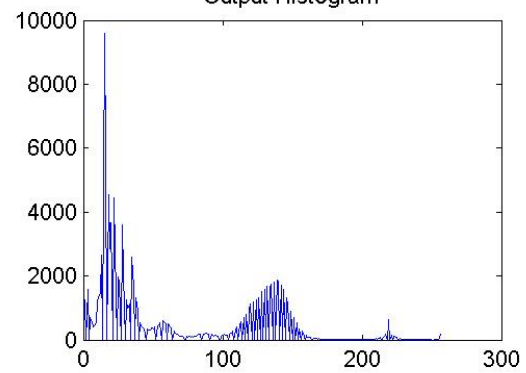
Original Histogram



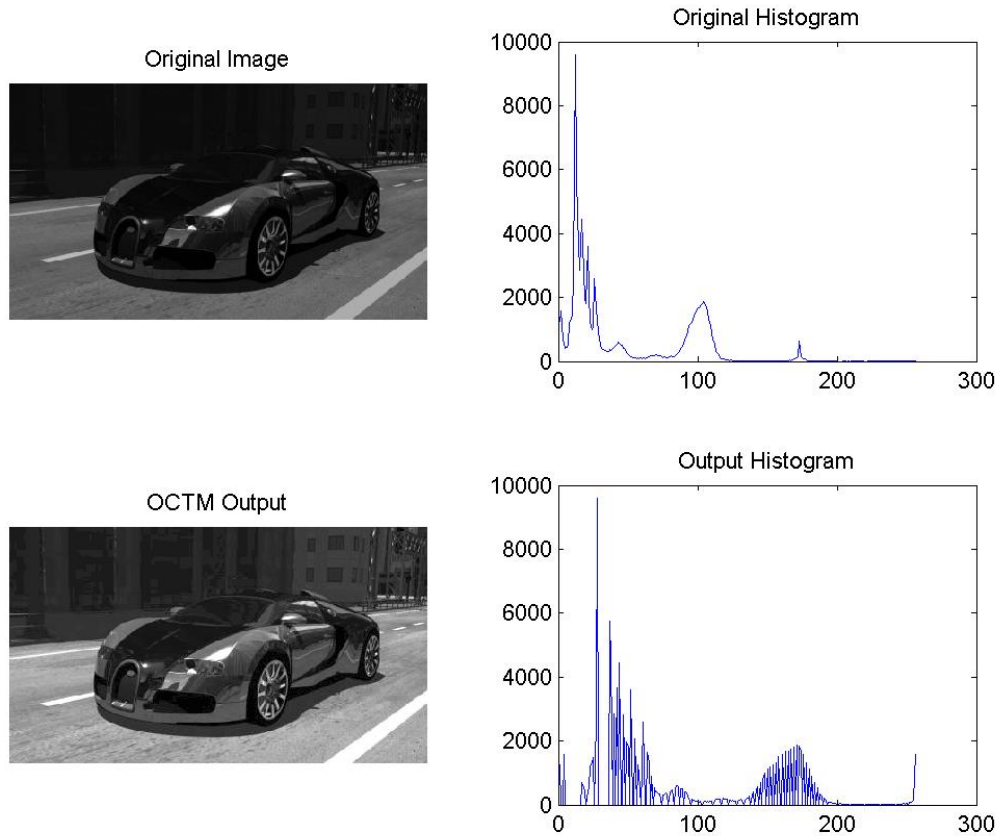
OCTM Output



Output Histogram

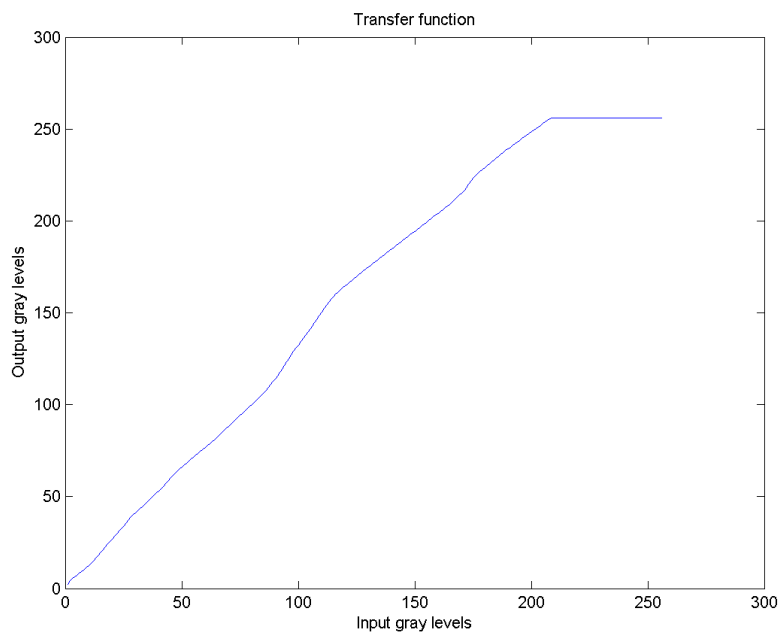


Test 3: $d = 3$, $\text{diff} = 2t$



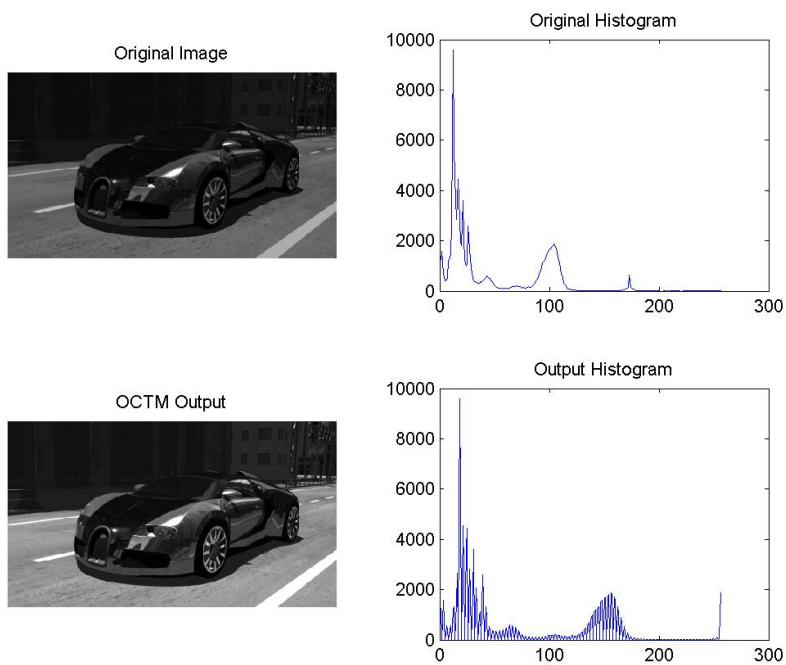
The differences between the images relative to the intensity range variable can be seen by the images above. While at $\text{diff} = 0$, the contrast of the image seemed to decrease, increasing the value seems to increase the contrast. $\text{Diff} = 1$ seems to be the best choice for this example since at $\text{diff} = 2$ the image may become too bright and slightly pixelated in the darker areas.

The transfer function at $\text{diff} = 1$ is shown here:



The value of the distortion that seemed to provide the best images was found to be 3 for this example. A different value of distortion is shown below.

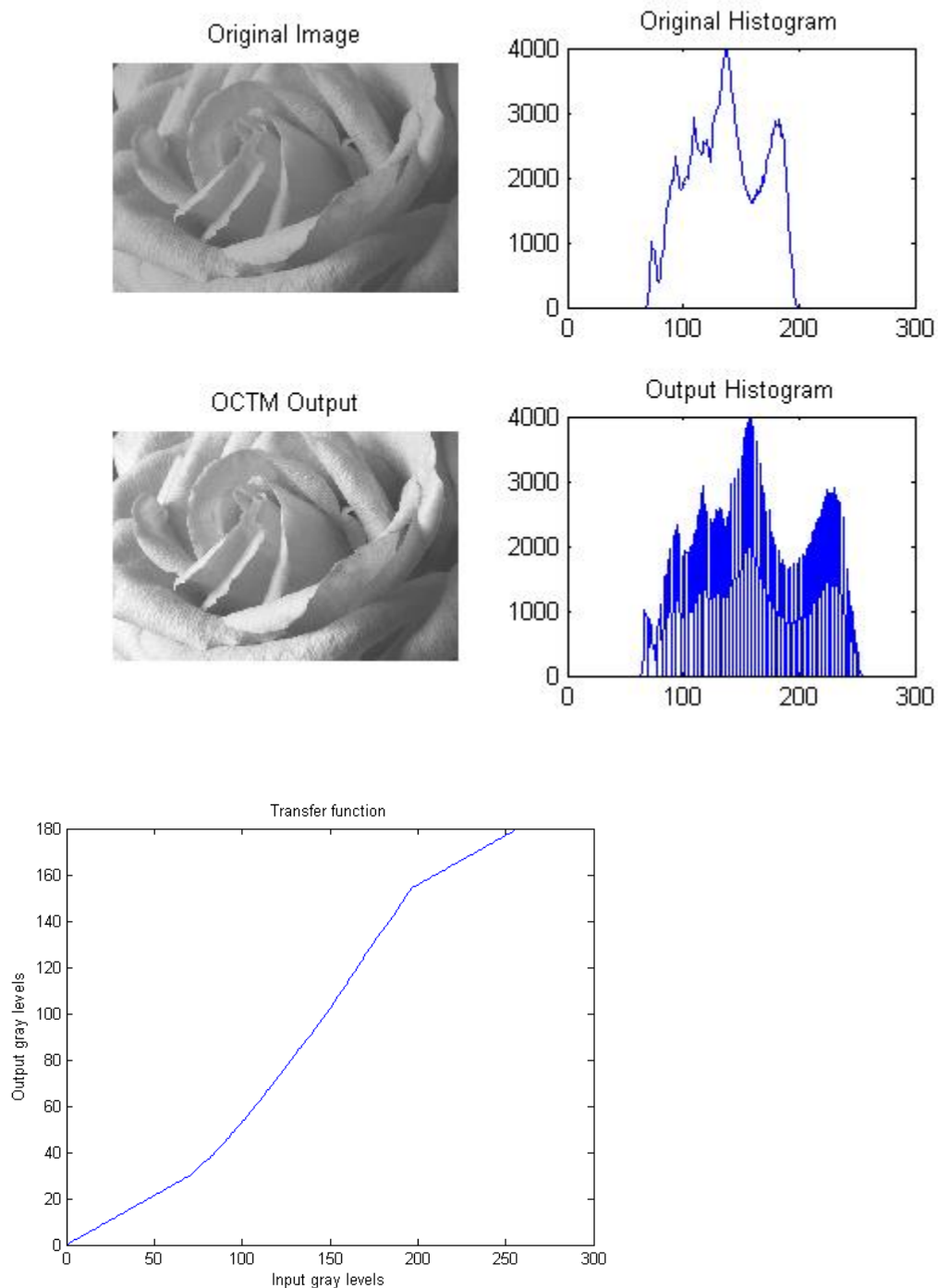
Test 4: $d = 2$, $\text{diff} = 1$



For each example of OCTM, the parameters can be adjusted to provide the best looking image. Therefore the image is highly dependent on these parameters for the constraint on the optimization of the contrast.

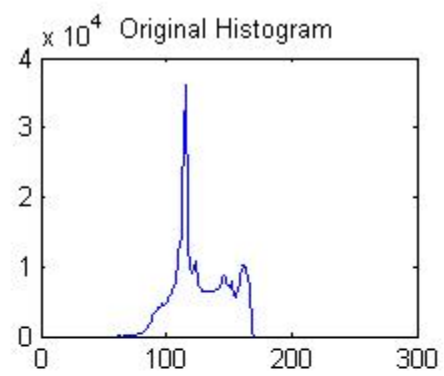
Here are a few more examples where the choice of parameters produced desirable results.

Test 5:

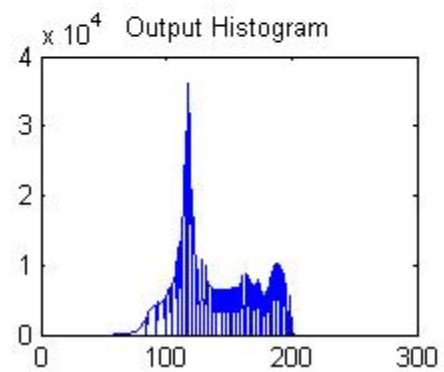


Test 6:

Original Image

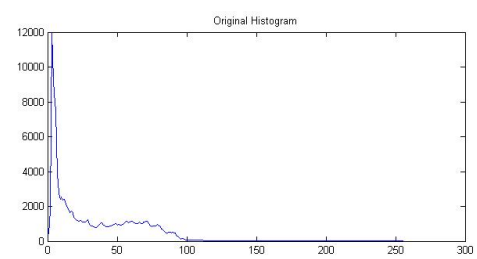


OCTM Output

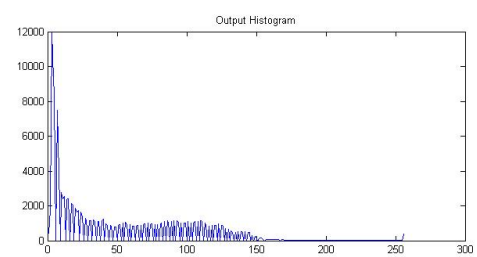


Test 7:

Original Image

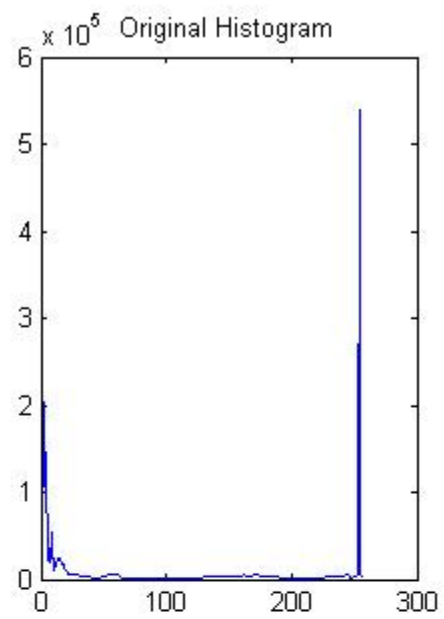


OCTM Output

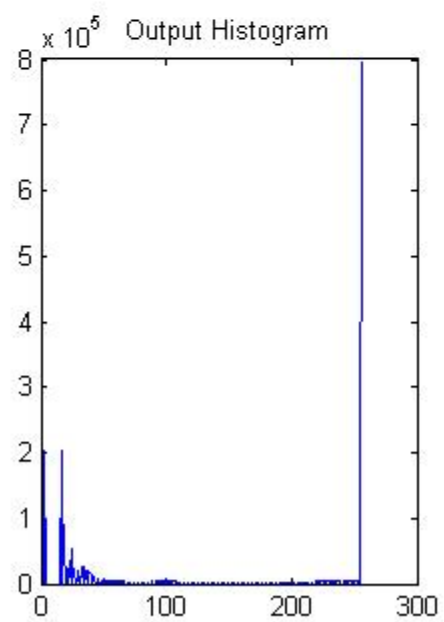


Test 8:

Original Image



OCTM Output



Matlab Code:

Histogram Equalization

```
clear all; close all; clc;

inputfile = 'low-contrast';
inputtext = '.png';
input = [inputfile,inputtext];
InputImage = imread(input);

[rows, columns, numberOfColorChannels] = size(InputImage);

% Convert to grayscale image
if numberOfColorChannels == 3
    InputImage = rgb2gray(InputImage);
end

pixelTotal=size(InputImage,1)*size(InputImage,2);

% Calculating number of pixels for each occurrence of a gray level
freq=zeros(256,1);

for i=1:size(InputImage,1)
    for j=1:size(InputImage,2)
        value=InputImage(i,j);
        freq(value+1)=freq(value+1)+1;
    end
end

sum=0;
no_bins=255;

% Cumulative distribution probability
cdf=zeros(256,1);
s=zeros(256,1);

for i=1:256
    sum=sum+freq(i);
    cdf(i)=sum;
    s(i)=round((cdf(i)/pixelTotal)*no_bins); %transfer function
end

equalizedImage=uint8(zeros(size(InputImage,1),size(InputImage,2)));

for i=1:size(InputImage,1)
    for j=1:size(InputImage,2)
        equalizedImage(i,j)=s(InputImage(i,j)+1);
        %mapping gray level of original image to equalized image
    end
end

name = ['histeq_',inputfile];
```

```

figure('Position', [100, 425, 1400, 400])
subplot(1,2,1);
imshow(InputImage);
title('Original Image');
subplot(1,2,2);
plot(imhist(InputImage));
title('Histogram of Original');
print([name, '_original'], '-dpng');
figure('Position', [100, 0, 1400, 400])
subplot(1,2,1);
imshow(equalizedImage);
title('Histogram equalization');
subplot(1,2,2);
plot(imhist(equalizedImage));
title('Histogram of Equalized Image');
print([name, '_enhanced'], '-dpng');

```

Optimal Contrast Tone Mapping

```

clc; clear all; close all;

inputfile = 'lcflower';
inputtext = '.jpg';
input = [inputfile, inputtext];
input_img = imread(input);
[rows, columns, numberOfColorChannels] = size(input_img);

if numberOfColorChannels == 3
    input_img = rgb2gray(input_img);
end

array_hist = imhist(input_img); % probability from the histogram

% constraint parameters
L = 256;
Lbar = 256;
[x,y] = size(input_img); %size of the input image is extracted
npixels = x*y; %the total number of pixels calculated
probability = array_hist/npixels; %probability of the occurrence of a pixel
diff = 1; % for average intensity constraint

% constraints
I=[1:L];

A = [ones(1,256) % fundamental constraints
     triu(repmat(probability',L,1))]; % average intensity constraint
% -triu(repmat(probability',L,1))];

b = [255 % fundamental constraints
     diff+probability.*I']; % average intensity constraint
% diff-probability.*I'];

% linprog arguments

```

```

Aeq = []; % empty matrix
Beq = []; % empty matrix
distortion = 2.5; % distortion factor
lower_bound = (1/distortion)*ones(1,L);
upper_bound = Lbar*ones(1,L);
% calculation of the step increment
Sj = linprog(-probability,A,b,Aeq,Beq,lower_bound,upper_bound);
output = input_img; % input image is set up as a new variable in order to be
used in the re allocation of pixels.
% for loop, used to re-allocate the pixels
for i=1:x
    for j=1:y
        output(i,j) = (sum(Sj(1:input_img(i,j)))+0.5);
    end
end

name =
['octm_avgint_',inputfile,'_d',num2str(distortion),'_diff',num2str(diff)];
figure(1)
subplot(2,2,1);
imshow(input_img);
title('Original Image');
subplot(2,2,2)
plot(imhist(input_img));
title('Original Histogram');
subplot(2,2,3);
imshow(output);
title('OCTM Output');
subplot(2,2,4);
plot(imhist(output));
title('Output Histogram');
% print(name,'-dpng');

% TRANSFER FUNCTION
% real one using Sj values
X=(1:L)';
Y=zeros(L,1);
for i=1:256
    Y(i,1)=sum(Sj(1:i));
end

figure(2)
plot(X,Y);
title('Transfer function');
xlabel('Input gray levels');
ylabel('Output gray levels');
% print([name,'_TRANS2'],'-dpng');

```