```
% Problem 1: Extraction Bitplanes from Image and Image Reconstruction from
% Bitplanes.
clear all;
%----Part 1 : Read the original image :slope.tif', which is an 8 bits/pixel gray scale \checkmark
slope image = imread('slope.tif'); % reading the image from an external source with the \checkmark
help of imread funciton.
%----Part 2 : Extract its 8 bitplanes as 8 binary images. storing image
%information in new variable
converted slope image = double(slope image); % converting each pixel of the image in \checkmark
double.
% extracting all bit one by one from 1st to 8th in variable from bitplane 1 toarksim
bitplane 8 respectively.
% We are converting all the images into binary format. That is why we have
% used mod here.
bit plane 8 = mod(converted slope image, 2); % Least Significant Bitplane
bit plane 7 = mod(floor(converted slope image/2), 2);
bit plane 6 = mod(floor(converted slope image/4), 2);
bit plane 5 = mod(floor(converted slope image/8),2);
bit plane 4 = mod(floor(converted slope image/16),2);
bit plane 3 = mod(floor(converted slope image/32),2);
bit plane 2 = mod(floor(converted slope image/64),2);
bit plane 1 = mod(floor(converted slope image/128),2); % Most Significant Bitplane
%imshow(a)
%----Part 3 : Reconstruct 8 versions of the original image.
% Reconstructing the image using the most significant (upper) bitplane only.
reconstructed_image_1 = (2 * (2 * (2 * (2 * (2 * (2 * bit_plane_1)))))); % 128 *
reconstructed image 1 = uint8(reconstructed image 1); % It will convert each and every ✓
pixel value of reconstructed image 1 into the range of 0 to 255.
%reconstructing the image using upper 2 bitplanes only
reconstructed image 2 = (2 * (2 * (2 * (2 * (2 * (2 * bit plane 1 + \checkmark)
bit plane 2))))))); % 128 * bitplane1 + 64 * bitplane2
reconstructed image 2 = uint8(reconstructed image 2); % It will convert each and every ✓
pixel value of reconstructed image 2 into the range of 0 to 255.
%reconstructing the image using upper 3 bitplanes only
reconstructed image 3 = (2 * (2 * (2 * (2 * (2 * (2 * bit plane 1 + bit plane 2) + \checkmark
bit plane 3)))))); % 128 * bitplane1 + 64 * bitplane2 + 32 * bitplane3
reconstructed image 3 = uint8(reconstructed image 3); % It will convert each and every ✓
pixel value of reconstructed image 3 into the range of 0 to 255.
%reconstructing the image using upper 4 bitplanes only
reconstructed image 4 = (2 * (2 * (2 * (2 * (2 * (2 * bit plane 1 + bit plane 2) + \checkmark
```

```
bit plane 3) + bit plane 4))))); % 128 * bitplane1 + 64 * bitplane2 + 32 * bitplane3 + ∠
16 * bitplane4
reconstructed image 4 = uint8(reconstructed image 4); % It will convert each and every ✓
pixel value of reconstructed image 4 into the range of 0 to 255.
%reconstructing the image using upper 5 bitplanes only
reconstructed image 5 = (2 * (2 * (2 * (2 * (2 * (2 * bit_plane_1 + bit_plane_2) + \checkmark)
bit plane 3) + bit plane 4) + bit plane 5) ))); % 128 * bitplane1 + 64 * bitplane2 + 32 ✓
* bitplane3 + 16 * bitplane4 + 8 * bitplane5
reconstructed image 5 = uint8(reconstructed image 5); % It will convert each and every ✓
pixel value of reconstructed image 5 into the range of 0 to 255.
%reconstructing the image using upper 6 bitplanes only
reconstructed image 6 = (2 * (2 * (2 * (2 * (2 * (2 * bit plane 1 + bit plane 2) + \checkmark
bit plane 3) + bit plane 4) + bit plane 5) + bit plane 6) )); % 128 * bitplane1 + 64 * ✔
bitplane2 + 32 * bitplane3 + 16 * bitplane4 + 8 * bitplane5 + 4 * bitplane6
reconstructed image 6 = uint8(reconstructed image 6); % It will convert each and every ✓
pixel value of reconstructed image 6 into the range of 0 to 255.
%reconstructing the image using upper 7 bitplanes only
reconstructed image 7 = (2 * (2 * (2 * (2 * (2 * (2 * bit plane 1 + bit plane 2) + \checkmark
bit plane 3) + bit plane 4) + bit plane 5) + bit plane 6) + bit plane 7) ); % 128 *\checkmark
bitplane1 + 64 * bitplane2 + 32 * bitplane3 + 16 * bitplane4 + 8 * bitplane5 + 4 * ✔
bitplane6 + 2 * bitplane7
reconstructed image 7 = uint8(reconstructed image 7); % It will convert each and every \checkmark
pixel value of reconstructed image 7 into the range of 0 to 255.
%reconstructing the image using upper 8 bitplanes only
reconstructed image 8 = (2 * (2 * (2 * (2 * (2 * (2 * bit plane 1 + bit plane 2) + \checkmark
bit plane 3) + bit plane 4) + bit plane 5) + bit plane 6) + bit plane 7) + \checkmark
bit plane 8); % 128 * bitplane1 + 64 * bitplane2 + 32 * bitplane3 + 16 * bitplane4 + 8 ✔
* bitplane5 + 4 * bitplane6 + 2 * bitplane7 + 1 * bitplane8
reconstructed image 8 = uint8(reconstructed image 8); % It will convert each and every ✓
pixel value of reconstructed image 8 into the range of 0 to 255.
%----Part 4: Create 4 figures, each with 4 sub-figures of 2x2 layout
% In Figure 1, show the upper 4 bitplanes in 4 sub-figures.
figure; % figure creates figure graphics objects. figure objects are the individual 🗹
windows on the screen in which MATLAB displays graphical output.
subplot(2,2,1) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=1.
imshow(bit plane 1) % It will display the gray-scale image in the figure.
title('Bitplane 1 image') % It will add the specified title for the current plot.
subplot(2,2,2) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=p=2.
imshow(bit plane 2) % It will display the gray-scale image in the figure.
title('Bitplane 2 image') % It will add the specified title for the current plot.
subplot(2,2,3) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=3.
imshow(bit plane 3) % It will display the gray-scale image in the figure.
```

```
subplot(2,2,4) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=4.
imshow(bit plane 4) % It will display the gray-scale image in the figure.
title('Bitplane 4 image') % It will add the specified title for the current plot.
% In Figure 2, show the lower 4 bitplanes in 4 sub-figures.
figure; % figure creates figure graphics objects. figure objects are the individual ✓
windows on the screen in which MATLAB displays graphical output.
subplot(2,2,1) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=1.
imshow(bit plane 5) % It will display the gray-scale image in the figure.
title('Bitplane 5 image') % It will add the specified title for the current plot.
subplot(2,2,2) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=p=2.
imshow(bit plane 6) % It will display the gray-scale image in the figure.
title('Bitplane 6 image') % It will add the specified title for the current plot.
subplot(2,2,3) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=3.
imshow(bit plane 7) % It will display the gray-scale image in the figure.
title('Bitplane 7 image') % It will add the specified title for the current plot.
subplot(2,2,4) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=4.
imshow(bit plane 8) % It will display the gray-scale image in the figure.
title('Bitplane 8 image') % It will add the specified title for the current plot.
\% Figure 3, show the first 4 reconstructed images in 4 sub-figures, corresponding to m{arksigma}
using the upper 1, 2, 3, and 4 bitplanes, respectively.
figure; % figure creates figure graphics objects. figure objects are the individual \( \sim \)
windows on the screen in which MATLAB displays graphical output.
subplot(2,2,1) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=1.
imshow(reconstructed image 1) % It will display the gray-scale image in the figure.
title('Reconstructed Image using bitplane 1') % It will add the specified title for the \checkmark
current plot.
subplot(2,2,2) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=p=2.
imshow(reconstructed image 2) % It will display the gray-scale image in the figure.
title('Reconstructed Image using bitplanes 1-2') % It will add the specified title for 
the current plot.
subplot(2,2,3) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=3.
imshow(reconstructed image 3) % It will display the gray-scale image in the figure.
title('Reconstructed Image using bitplanes 1-3') % It will add the specified title for \checkmark
the current plot.
subplot(2,2,4) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the position specified by p. Here, m=n=2, p=4.
imshow(reconstructed image 4) % It will display the gray-scale image in the figure.
title('Reconstructed Image using bitplanes 1-4') % It will add the specified title for \checkmark
the current plot.
```

title('Bitplane 3 image') % It will add the specified title for the current plot.

% In Figure 4, show the remaining 4 reconstructed images in 4 sub-figures, \checkmark corresponding to using the upper 5, 6, 7 and 8 bitplanes, respectively.

figure; % figure creates figure graphics objects. figure objects are the individual \checkmark windows on the screen in which MATLAB displays graphical output.

subplot(2,2,1) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark creates axes in the position specified by p. Here, m=n=2, p=1.

imshow(reconstructed_image_5) % It will display the gray-scale image in the figure. title('Reconstructed Image using bitplanes 1-5') % It will add the specified title for \checkmark the current plot.

subplot(2,2,2) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark creates axes in the position specified by p. Here, m=n=p=2.

imshow(reconstructed_image_6) % It will display the gray-scale image in the figure. title('Reconstructed Image using bitplanes 1-6') % It will add the specified title for \checkmark the current plot.

subplot(2,2,3) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark creates axes in the position specified by p. Here, m=n=2, p=3.

imshow(reconstructed_image_7) % It will display the gray-scale image in the figure. title('Reconstructed Image using bitplanes 1-7') % It will add the specified title for \checkmark the current plot.

subplot(2,2,4) % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark creates axes in the position specified by p. Here, m=n=2, p=4.

imshow(reconstructed_image_8) % It will display the gray-scale image in the figure. title('Reconstructed Image using bitplanes 1-8') % It will add the specified title for \checkmark the current plot.

Figure 1:

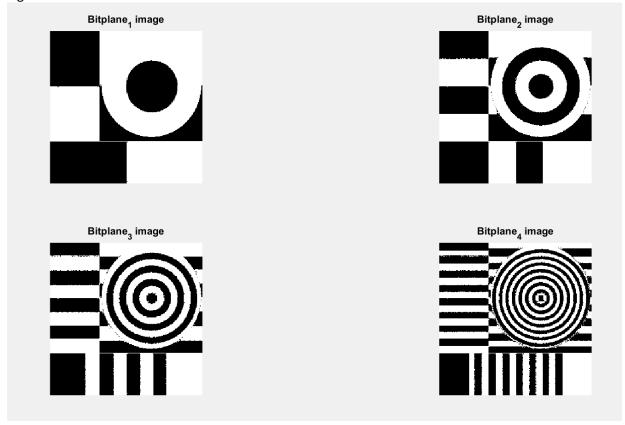
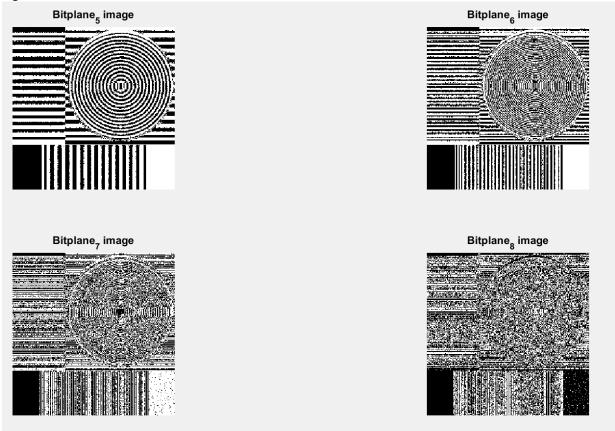


Figure 2:



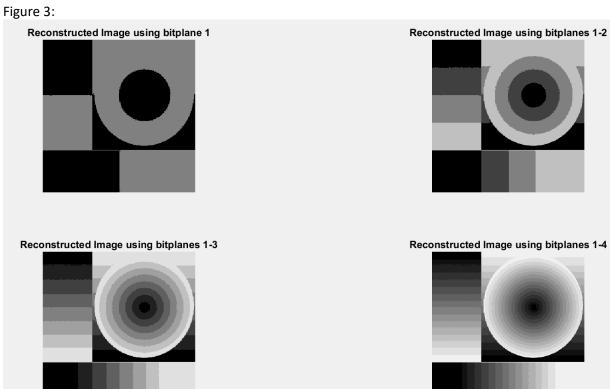
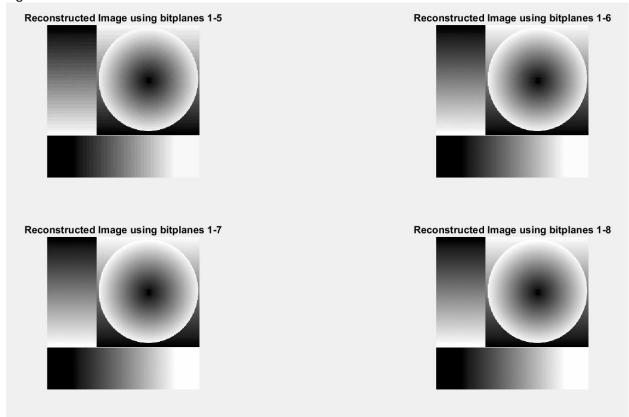


Figure 4:



Q1 Observation:

The original image is broken into multiple bitplanes (8 in all) in Figures 1 and 2, creating the appearance that the image has been fractured into 8 layers. After we recreated the image from these different bitplanes one at a time, two at a time, and so on, we can see the picture recovering its prior look bitplane by bitplane. The picture takes almost all of its shape in the first four reconstructed photographs in Figure 3, while the remaining rebuilt photos in Figure 4 are merely adding to the finer definitions. Because the first four Bitplanes 1-4 are the most significant, while Bitplanes 5-8 are the least crucial, we observe this. As demonstrated in the reconstruction, the upper bitplanes contain the bulk of the structural data, while the lower bitplanes contribute the distortions and gray areas. The reconstruction shows that a 6 bitplane reconstruction can almost precisely match the original image.

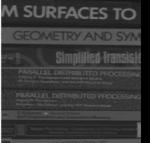
```
% Problem 2: Intensity Transformations: Gamma Mapping, Full-scale Contrast
% Stretch, and Histogram Equalization.
clear all;
input image = double(imread('books.tif')); % reading the original image. This image is \checkmark
8 bits/pixel gray-scale
% image.
% Plotting the original image.
figure % figure creates figure graphics objects. figure objects are the individual \checkmark
windows on the screen
% in which MATLAB displays graphical output.
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
imshow(uint8(input image)); % It will display the gray-scale image in the figure and
st it will convert each and every pixel value of the input image into the range of 0 tooldsymbolarksim
255.
title('Original Books Image'); % It will add the specified title for the current plot.
\max_{\alpha} = \max_{\alpha} (input_{\max}(:)); % It will calculate the maximum value from the input <math>\checkmark
% will be further used for image scaling for generating gamma mapped image.
min val = min(input image(:)); \% It will \bigcirc alare the minimum value from the input\checkmark
% will be further used for image scaling for generating gamma mapped image.
input image scaled = (input image - min val) / (max val - min val); % Here, we are \checkmark
scalling the pixel
% intensity. It is basically called the normalixation of all the pixel
% values in the image. We are scalling this image from [0, 255] to [0, 1]
% to apply power law.
gamma transformed image = (input image scaled.^{\circ}0.5) * 255; \% Here, we applied power law\checkmark
with gamma = 0.5 and
st after that we again need to rescale the image from [0,1] to [0, 255].Now, our gammam{arksigma}
mapped image is ready.
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and ✓
creates axes in the
% position specified by p. Here, m=n=p=2.
imshow(uint8(gamma transformed image)); % It will display the gray-scale image in the ∠
\$ it will convert each and every pixel value of the input image into the range of 0 tooldsymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksy
title('Gamma Transformed Image'); % It will add the specified title for the current\checkmark
plot.
full scale contrast stretched image = round((2^8-1) * input image scaled); % Here, we ✓
created the full scale
% contrast stretched image from the original input scaled image.
subplot(2,2,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
```

```
creates axes in the
% position specified by p. Here, m=n=2, p=3.
imshow(uint8(full scale contrast stretched image)); % It will display the gray-scale ✓
image in the figure and
\% it will convert each and every pixel value of the input image into the range of 0 to \checkmark
255.
title('Full Scale Contrast Streched Transformed Image'); % It will add the specified ✓
title for the current
% plot.
cummuluative histogram equalized image = cumsum(hist(reshape(input image,[],1), 🗸
linspace(0,255,256))); % It will
% create a cummulative histogram equalized image. hist() function will
% create a histogram bar chart of the elements in the given vector.
% reshape() method will change the shape of our image.
intermediate image = cummuluative histogram equalized image(uint8(input image) + 1); % ✓
It will create an
% intermediate image to create final histogram equalized image. uint8() will convert \swarrow
the value
% of each pixel from 0 to 255.
hei = round((2^8 - 1)*((intermediate image - min(intermediate image(:)))/(max ✓
(intermediate image(:)) - ...
    \min(\text{intermediate\_image(:))}))); % It will simply do the scalling of the image to \checkmark
generate the final histogram
% equalized image.
subplot(2,2,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=4.
imshow(uint8(hei)); % It will display the gray-scale image in the figure and
st it will convert each and every pixel value of the input image into the range of 0 tom{arksigma}
title('Histogram Equalized Image'); % It will add the specified title for the current
% plot.
figure; % figure creates figure graphics objects. figure objects are the individual \( \mathbf{L} \)
windows on the screen
% in which MATLAB displays graphical output.
original image histogram = hist(reshape(input image,[],1), linspace(0,255,256));
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
bar(original image histogram); % It will create a bar chart of the given image.
title('Histogram of Original Books Image'); % It will add the specified title for the
current
% plot.
gamma transformed image histogram = hist(reshape(gamma transformed image,[],1),\checkmark
linspace(0, 255, 256);
```

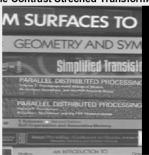
```
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and ✓
creates axes in the
% position specified by p. Here, m=n=p=2.
bar(gamma transformed image histogram); % It will create a bar chart of the given ✓
image.
title('Histogram of Gamma Transformed Image'); % It will add the specified title for ✓
the current
% plot.
fscs image histogram = hist(reshape(full scale contrast stretched image,[],1), linspace ✓
(0,255,256));
subplot(2,2,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=3.
bar(fscs image histogram); % It will create a bar chart of the given image.
title('Histogram of Full Scale Contrast Streched Image'); % It will add the specified
title for the current
% plot.
hei image histogram = hist(reshape(hei,[],1), linspace(0,255,256));
subplot(2,2,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=4.
bar(hei image histogram); % It will create a bar chart of the given image.
title('Histogram of Histogram Equalized Image'); % It will add the specified title for ✓
the current
% plot.
```

Figure 1:











Histogram Equalized Image

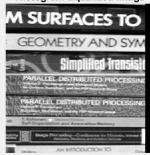
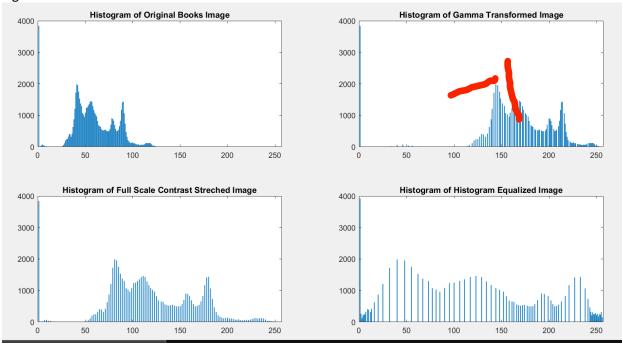


Figure 2:



Q2 Observation:

The image sharpened as we moved from Gamma intensity change to full-scale contrast stretch to histogram equalization. The histograms also revealed that the pixel intensities in the histogram equalized image are more balanced than the original image, implying that the histogram equalized image is the clearest of the four images. After applying the gamma rule, the pixel values (intensity) moved towards the center, slightly increasing the brightness. The contrast and brightness of the full-size extended image are higher, and the histogram reveals more uniformly distributed pixel values.

```
% Problem 3: Frequency Domain Low-pass, Band-pass and High-pass Filtering.
clear all;
% functionality of a particular functions.
% fft2() - It will return the two-dimensional fourier transform of a
% matrix using the fast fourier trtansform.
% fftshift() - It will rearrange a Fourier transform x by shifting the
% zero-frequency component to the center of the array.
% ifft2() - It will return the two-dimensional inverse fourier transform of a
% matrix using the fast fourier trtansform.
input image = double(imread("bridge.tif")); % reading the original image. This image is ✓
8 bits/pixel gray-scale
% image.
[n rows, n cols] = size(input image); % It will return a x * y dimension of the image. \checkmark
x represents the number of
% rows in the image and y represents the number of columns of a matrix formed from the arksim \prime
input image.
% for example if we have a 3-by-4 image then size function will return [3 4].
% The 2-D DFT of the original image and shifting the DC component.
input image 2ddft = fft2(input image); % 2D-DFT of the original image.
input image 2ddft shift = fftshift(input image 2ddft); % Shifting of the DC component.
%center of image
mid value = n rows/2; % finding the middle of the image.
% We want to do the indexing of the matrix, so we will create rectuangluar
% structures from the given array. It is basically done with the help of
% meshgrid function.
[vector col, vector row] = meshgrid(1:n cols, 1:n rows); % It will return 2-D grid ✓
coordinates based on the
% coordinates in vectors (1:n row) and (1:n col).
% Evaluating distance over the grid. Calculating the euclidean distance(distance \checkmark
between two points).
distance = ((vector row-mid value).^2+(vector col-mid value).^2).^0.5; % simply ✓
calculating the distance between
% two points.
% Here, we will design 3 filters in total. 1] Low-pass filter, 2] high-pass
% filter, and 3] bandpass filter.
% Designing low pass filter
distance low = (1/8) * mid value; % It is given in the problem statement that \checkmark
frequency response of the ideal
\$ low pass filter is equal to one-eighth of the distance from the center to the m{arkappa}
horizontal or vertical edge.
low pass filter = (distance <= distance low); % here the distance must be less than or \checkmark
```

```
equal to the frequency
% response of the ideal low pass filter.
low pass filtered image = input image 2ddft shift.*low pass filter; % Here, the '.'┕
after input image 2ddft shift
st indicates that it will take all the pixel values. i.e. it will take all the values of m{arksigma}
each and every rows and
st columns from the matrix. Also, we are applying a low pass filter on the shifted oldsymbolarksim
image.
low pass filtered spatial image = real(ifft2(fftshift(low pass filtered image))); % ✓
Here we are applying inverse
\$ 2D-DFT on the low pass filtered image to obtain the image in spatial domain. \$ So, \checkmark
first we apply fftshift and
% then ifft2 function for inverse 2D-DFT.
% Designing high pass filter
distance high = (1/2) * mid value; % It is given in the problem statement that \checkmark
frequency response of the ideal
% high pass filter is equal to one-twoth of the distance from the center to the oldsymbol{arkappa}
horizontal or vertical edge.
high pass filter = (distance >= distance high); % here the distance must be greater \checkmark
than or equal to the frequency
% response of the ideal high pass filter.
high pass filtered image = input image 2ddft shift.*high pass filter; % Here, the '.'┕
after input image 2ddft shift
\$ indicates that it will take all the pixel values. i.e. it will take all the values of \checkmark
each and every rows and
st columns from the matrix. Also, we are applying a high pass filter on the shifted m{arksigma}
image.
high pass filtered spatial image = real(ifft2(fftshift(high pass filtered image))); % ✓
Here we are applying inverse
% 2D-DFT on the high pass filtered image to obtain the image in spatial domain. % So, \checkmark
first we apply fftshift and
% then ifft2 function for inverse 2D-DFT.
% Deigning band pass filter
band pass filter = (distance >= distance low & distance <= distance high); % here the ✓
distance must be greater
\$ than or equal to the frequency response of the ideal low pass filter and less than or m{arkappa}
equal to the frequency
% response of the ideal high pass filter.
band pass filtered image = input image 2ddft shift.*band pass filter; % Here, the '.'\(\varphi\)
after input image 2ddft shift
\$ indicates that it will take all the pixel values. i.e. it will take all the values of \checkmark
each and every rows and
\$ columns from the matrix. Also, we are applying a band pass filter on the shifted oldsymbol{arkappa}
image.
```

band pass filtered spatial image = real(ifft2(fftshift(band pass filtered image))); % 🗸

Here we are applying inverse

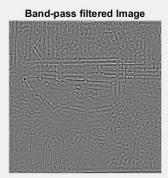
```
% 2D-DFT on the band pass filtered image to obtain the image in spatial domain. % So, \checkmark
first we apply fftshift and
% then ifft2 function for inverse 2D-DFT.
f = figure; % figure creates figure graphics objects. figure objects are the individual \checkmark
windows on the screen
% in which MATLAB displays graphical output.
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
imshow(uint8(input image)); % It will display the gray-scale image in the figure and
st it will convert each and every pixel value of the input image into the range of 0 tom{arkappa}
255.
title('Original Image'); % It will add the specified title for the current plot.
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=p=2.
imshow(low pass filtered spatial image,[]); % It will display the gray-scale image in ✓
the figure.
title('Low-pass filtered Image'); % It will add the specified title for the current ✓
plot.
subplot(2,2,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=3.
imshow(band pass filtered spatial image, []); % It will display the gray-scale image in ✓
the figure.
title('Band-pass filtered Image'); % It will add the specified title for the current ✓
plot.
subplot(2,2,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=4.
imshow(high pass filtered spatial image, []); % It will display the gray-scale image in ✓
title('High-pass filtered Image'); % It will add the specified title for the current ✓
plot.
f = figure; % figure creates figure graphacs objects. figure objects are the individual \( \subseteq \)
windows on the screen
% in which MATLAB displays graphical outpot.
\max\_amplitude = \max(input\_image(:)); % It will simply find the maximum amplitude of the <math>\checkmark
log shifted img = \log(1+abs(input image 2ddft shift)); % Apply log transform on the \checkmark
2ddft shifted image
% and make it log shifted image.
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
imshow(log shifted img./max amplitude,[]); % It will display the gray-scale image in ✓
the figure.
title('Original Image'); % It will add the specified title for the current plot.
```

```
log shifted low = \log(1+abs(low pass filtered image)); % Apply log transform on the <math>\checkmark
2ddft shifted image
% and make it log shifted image.
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=p=2.
imshow(log_shifted_low./max_amplitude, []); % It will display the gray-scale image in ✓
the figure.
title('Low-pass filtered'); % It will add the specified title for the current plot.
log shifted band = \log(1+abs) (band pass filtered image)); % Apply log transform on the \checkmark
2ddft shifted image
% and make it log shifted image.
subplot(2,2,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=3.
imshow(log shifted band./max amplitude, []); % It will display the gray-scale image in ✓
title('Band-pass filtered'); % It will add the specified title for the current plot.
log shifted high = \log(1+abs) (high pass filtered image)); % Apply log transform on the \checkmark
2ddft shifted image
% and make it log shifted image.
subplot(2,2,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=4.
imshow(log shifted high./max amplitude, []); % It will display the gray-scale image in ✓
the figure.
```

title('High-pass filtered'); % It will add the specified title for the current plot.

Figure 1:







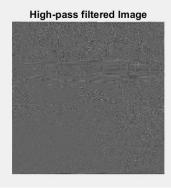
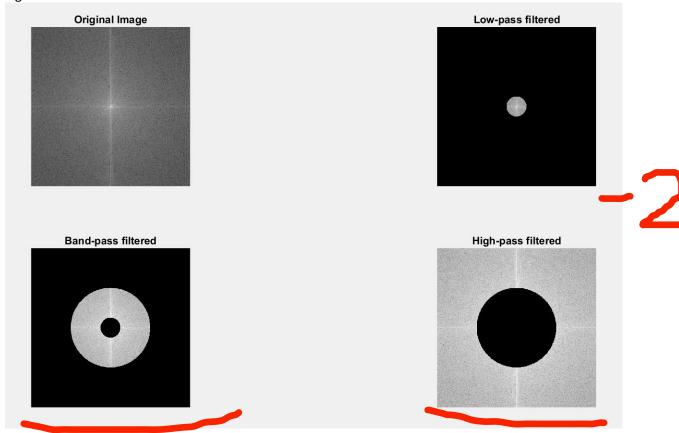


Figure 2:



Q3 Observation:

The low pass filter smoothens out the images by reducing the high-intensity pixels while leaving the low-intensity pixels alone. As can be seen from the amplitude response, pixels with a percent higher intensity than the cut-off value are masked out. Similarly, the high pass filter sharpens the image. By suppressing low Pixel intensities and reducing the percent intensity, the band pass filter improves the edges while minimizing noise. Extremely high pixel intensities smoothens out the image by filtering out all of the high frequency components after sending it through a low pass filter. The image is stripped of all low frequency components after going through a high pass filter, leaving just the edges.

```
% Problem 4 - Image Deblurring by Frequency Domain Inverse Filter Design.
clear all;
% functionality of a particular functions.
% fft2() - It will return the two-dimensional fourier transform of a
% matrix using the fast fourier trtansform.
% fftshift() - It will rearrange a Fourier transform x by shifting the
% zero-frequency component to the center of the array.
% ifft2() - It will return the two-dimensional inverse fourier transform of a
% matrix using the fast fourier transform.
% mesh() - It will create a 3D surface that has a sloid edge colors and no face colors.
input image = double(imread("text.tif")); % reading the original image. This image is 8 ✓
bits/pixel gray-scale
size of filter = 21; % We are simply defining the filter size and it is used further in \checkmark
the below code.
standard deviation = 1; % We are defining the value of standard deviation as per the \checkmark
given problem.
[vector col, vector row] = meshgrid(-floor(size of filter/2):floor(size of filter/2), -✓
    floor(size of filter/2):floor(size of filter/2)); % It will return 2-D grid ✓
coordinates based on the
st coordinates in vectors. This basically means that we are using the cropped image of m{arkappa}
the central part, so that
% input image and the output image, bith has the same size.
gaussian filter = \exp(-(\text{vector col.}^2 + \text{vector row.}^2))/(2*\text{standard deviation}^2))/\checkmark
(2*pi*standard deviation^2);
% Here, we are just
\$ normilizing the gaussian filter such that the sum of all the coefficients will bem{arksigma}
equal to 1. This is given in
% the problem statement. We are just normalizing the gaussian filter.
blurring filter = gaussian filter./sum(gaussian filter(:)); % created a blurring filter ✓
which will be used
% to give blurring effect to an image.
blurred image = conv2(input image, blurring filter, 'same'); % we applied 2D ✓
convolution between the original
\$ input image and the gaussian filter to get a blurred image as a resulting image.oldsymbol{arkappa}
Hence, we have applied
% conv2() function on the input image.
% Frequency domain of the gaussian filter.
gaussian filter freq domain = fft2(blurring filter); % Here, we are applying 2D-DFT to \checkmark
the spatial domain
\$ Gaussian filter to obtain frequency domain gaussian filter. Hence, we have applied oldsymbol{arepsilon}
fft2() function on the
% blurred image.
```

```
%Inverse Gaussian Filter
freq domain inverse gaussian filter = 1./gaussian filter freq domain; % Here, we have \checkmark
created a frequency
\$ domain inverse gaussian filter by simply taking reciprocal value of each and everyoldsymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymb
pixel value of frequency
\$ domain Gaussian filter coefficients. The '.'indicates that we have taken all the m{arkappa}
values from the image matrix.
deblurring spatial domain inverse filter = real(ifft2 ✓
(freq domain inverse gaussian filter)); % Here, as per the
\$ problem statement we have applied the inverse 2D-DFT to generate spatial domainm{arkappa}
inverse gaussian filter. For
% this we have implemented ifft2() method.
% convolution between the blurred image and the deblurred image obtained from spatial \checkmark
domain inverse filter.
deblurred image = conv2(blurred image, deblurring spatial domain inverse filter, ✓
'same'); % we applied 2D
st convolution between the spatial domain gaussian filtered image and the spatial domain oldsymbolarksim
inverse gaussian filter
\$ to get a spatial domain blurred image as a resulting image. Hence, we have applied oldsymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksymbolarksym
conv2() function on the
% input image.
% Intensity transformation and DC shift for corresponding images.
original image fft = fft2(input image); % Here, we are applying 2D-DFT to the input \checkmark
image to obtain
% fourier transformed image. Hence, we have applied fft2() function on the input arksim \prime
original image intensity transform = log(1 + abs(original image fft)); % Apply log ✓
transform on the 2ddft
% image and make it log shifted image.
original image frequency = fftshift(original image intensity transform./ ...
          max(original image intensity transform(:))); % DC shift for the corresponding ✓
images.
blurred image fft = fft2(blurred image); % Here, we are applying 2D-DFT to the blurred \checkmark
image to obtain
% fourier transformed image. Hence, we have applied fft2() function on the blurred arksim \prime
blurred image intensity transform = log(1 + abs(blurred image fft)); % Apply log ✓
transform on the 2ddft
% image and make it log shifted image.
blurred image frequency = fftshift(blurred image intensity transform./max ✓
(blurred image intensity transform(:)));
% DC shift for the corresponding images.
deblurred image fft = fft2(deblurred image); % Here, we are applying 2D-DFT to the \checkmark
deblurred image to obtain
% fourier transformed image. Hence, we have applied fft2() function on the deblurred \checkmark
image.
```

creates axes in the

```
deblurred image intensity transform = log(1 + abs(deblurred image fft)); % Apply log ✓
transform on the 2ddft
% image and make it log shifted image.
deblurred image frequency = fftshift(deblurred image intensity transform./ ...
    max(deblurred image intensity transform(:))); % DC shift for the corresponding ✓
images.
% Plot for images and figures
figure % figure creates figure graphics objects. figure objects are the individual \checkmark
windows on the screen
% in which MATLAB displays graphical output.
subplot(2,3,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=2, n=3, p=1.
imshow(uint8(input image)); % It will display the gray-scale image in the figure and
\$ it will convert each and every pixel value of the input image into the range of 0 to m{arkappa}
255.
title('Original Text Image'); % It will add the specified title for the current plot.
subplot(2,3,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=2, n=3, p=2.
imshow(uint8(blurred image)); % It will display the gray-scale image in the figure and
st it will convert each and every pixel value of the input image into the range of 0 tom{arkappa}
255.
title('Blurred Text Image'); % It will add the specified title for the current plot.
subplot(2,3,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=2, n=p=3.
imshow(uint8(deblurred image)); % It will display the gray-scale image in the figure ✓
st it will convert each and every pixel value of the input image into the range of 0 toarksim
title('Deblurred Text Image'); % It will add the specified title for the current plot.
subplot(2,3,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=2, n=3, p=4.
imshow(original image frequency, []); % It will display the gray-scale image in the ✓
figure and
st it will convert each and every pixel value of the input image into the range of 0 tooldsymbolarksim
title('Original Image Frequency'); % It will add the specified title for the current \checkmark
plot.
subplot(2,3,5); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
```

```
% position specified by p. Here, m=2, n=3, p=5.
imshow(blurred image frequency,[]); % It will display the gray-scale image in the ∠
figure and
st it will convert each and every pixel value of the input image into the range of 0 tooldsymbolarksim
255.
title('Blurred Image Frequency'); % It will add the specified title for the current ✓
plot.
subplot(2,3,6); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=2, n=3, p=6.
imshow(deblurred image frequency,[]); % It will display the gray-scale image in the \checkmark
figure and
% it will convert each and every pixel value of the input image into the range of 0 to m{arkappa}
title('Deblurred Image Frequency'); % It will add the specified title for the current 
plot.
응응응응용
figure % figure creates figure graphics objects. figure objects are the individual 🗹
windows on the screen
% in which MATLAB displays graphical output.
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
mesh(blurring filter); % It will create a 3D surface that has a sloid edge colors and \checkmark
no face colors.
title('Spatial domain Gaussian Blur Filter'); % It will add the specified title for the
current plot.
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and ✓
creates axes in the
% position specified by p. Here, m=n=p=2.
mesh(abs(fftshift(fft2(blurring filter)))); % It will create a 3D surface that has a ✓
sloid edge colors and no
% face colors.
title('Magnitude of Fourier Domain GBF'); % It will add the specified title for the
current plot.
subplot(2,2,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=3.
mesh (deblurring spatial domain inverse filter); \% It will create a 3D surface that has \checkmark
a sloid edge colors and no
% face colors.
title('Spatial domain Gaussian Deblur Filter'); % It will add the specified title for \checkmark
the current plot.
```

 $\operatorname{subplot}(2,2,4)$; % $\operatorname{subplot}(m,\ n,\ p)$ divides the current figure into an m-by-n grid and \mathbf{k}' creates axes in the

% position specified by p. Here, m=n=2, p=4.

mesh(abs(fftshift(freq_domain_inverse_gaussian_filter))); % It will create a 3D surface \checkmark that has a sloid edge colors and

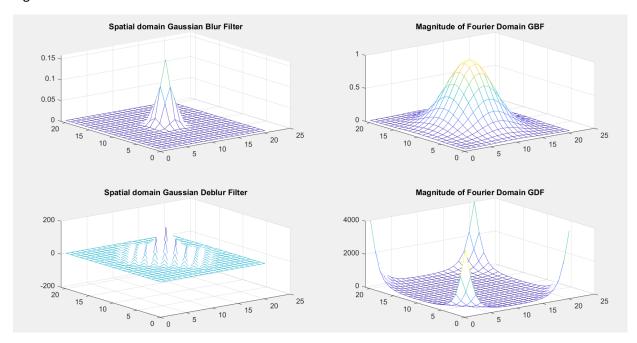
% no face colors.

title('Magnitude of Fourier Domain GDF'); % It will add the specified title for the \checkmark current plot.

Figure 1:

Original Text Image MT-LEVEL Sprinet3/home/u/rjkroeger/vf; Espinet3/home/u/rjkroeger/vf; Espine

Figure 2:



Q4 Observation:

The blurring effect on the picture was caused by Gaussian noise. Darker points were found farther distant from the brightest point in the middle of the blurred picture's amplitude spectrum. When the original and deblurred images amplitude spectrums are examined, we can see that all of the spots are equally intense and similar in both. A low-pass filter and the Gaussian filter blurs an image (smoothening). The deblurred picture is almost similar to the original, with the high frequency components enhanced for a little gain in sharpness.

```
% Problem 5: Image Interpolation: Nearest Neighbor and Bilinear Interpolation
clear all;
input image = imread('einstein.tif'); % reading the original image. This image is 8 ¥
bits/pixel gray-scale
% image.
size of image = size(input image); % It will simply calculate the size of the image.
downsampling factor = 3; % we are just defining the downsampling factor. It is as per\checkmark
given in the problem
% statement.
% Downsampling the image in horizontal and vertical directions
downsampled image d3 = input image((downsampling factor+1)/2:downsampling factor:end, \checkmark
    (downsampling factor+1)/2:downsampling factor:end); % it will downsample the image ✓
by a factor of D in
\% horizontal and vertical directions, where the top-left sample should be at the oldsymbol{arkappa}
location of ((D+1)/2, (D+1)/2)
% in the original image.
% upsampling
upsampled image d3 = zeros(size of image); % It will simply upsample the image. ✓
Upsampling of the downsampled
\$ image to have exactly the size of the original image by filling zeros in the missing m{arkappa}
pixels.
p = 1; % initializing the value of p.
for i = (downsampling factor+1)/2:downsampling factor:size of image % for loop as per ✓
the given condition.
    q = 1; % initializing the value of q.
    for j = (downsampling factor+1)/2:downsampling factor:size of image
    upsampled image d3(i,j) = downsampled image <math>d3(p,q); % for loop as per the given \checkmark
condition.
        q = q + 1; % incrementing the value of q.
    end % end for loop.
    p = p + 1; % incrementing the value of q.
end % end for loop.
% Convolution with block shape for D=3
block filter d3 = [1 1 1]; % creating the block filter of shape and size D.
block conv image d3 = upsampled image d3; % the block convolutional filter shold be of \checkmark
the same size as of the
% upsampled image.
for i = 1:size of image % for loop condition
block conv image d3(i,:) = conv(block conv image d3(i,:), block filter d3, 'same'); % <
Here, the convolutions
% of the rows of the matrix. convolution operation for rows.
end % for loop ends
```

```
for j = 1:size of image % for loop condition
block conv image d3(:,j) = conv(block conv image d3(:,j), block filter d3.', 'same'); % 🗸
Here, the convolutions
% of the rows of the matrix. convolution operation for columns.
end % for loop ends
% Convolution with triangle shape for D=3
triangle filter d3 = (1/3) * [1 2 3 2 1]; % defining the filter as per the given \checkmark
problem.
triangle_conv_image_d3 = upsampled_image d3; % the block convolutional filter shold be 2
of the same size as of the
% upsampled image.
for i = 1:size of image % for loop condition
    triangle conv image d3(i,:) = conv(triangle conv image d3(i,:), \checkmark
triangle filter d3,'same'); % Here, the
    % convolutions of the rows of the matrix. convolution operation for rows.
end % for loop ends
for j = 1:size of image % for loop condition
    triangle_conv_image_d3(:,j) = conv(triangle conv image d3(:,j), 
triangle filter d3.', 'same'); % Here, the
    % convolutions of the rows of the matrix. convolution operation for columns.
end % for loop ends
% Figure plotting for D=3
figure; % figure creates figure graphics objects. figure objects are the individual \checkmark
windows on the screen
% in which MATLAB displays graphical output.
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
imshow(input image); % It will display the gray-scale image in the figure.
title('Original image'); % It will add the specified title for the current plot.
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=p=2.
imshow(uint8(upsampled image d3)); % It will display the gray-scale image in the figure ✓
and
\$ it will convert each and every pixel value of the input image into the range of 0 to {m \ell}'
255.
title('Upsampled image (D=3)'); % It will add the specified title for the current plot.
subplot(2,2,3); % subplot(m, n, p) divides the current figure into an m-by-n grid and ✓
creates axes in the
% position specified by p. Here, m=n=2, p=3.
imshow(uint8(block conv image d3)); % It will display the gray-scale image in the ✓
figure and
% it will convert each and every pixel value of the input image into the range of 0 to \checkmark
title('Interpolated image using Nearest Neighbour (D=3)'); % It will add the specified ✓
```

```
title for the current plot.
subplot(2,2,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=4.
imshow(uint8(triangle conv image d3)); % It will display the gray-scale image in the ✓
figure and
st it will convert each and every pixel value of the input image into the range of 0 tooldsymbol{arkappa}
255.
title('Image using Bilinear interpolation (D=3)'); % It will add the specified title
for the current plot.
% D = 7
downsampling factor = 7; % we are just defining the downsampling factor. It is as per \checkmark
given in the problem
% statement.
%Downsampling the image in horizontal and vertical directions
downsampled image d7 = input image((downsampling factor+1)/2:downsampling factor:end, \checkmark
    (downsampling factor+1)/2:downsampling factor:end); % it will downsample the image ✓
by a factor of D in
% horizontal and vertical directions, where the top-left sample should be at the m{arksigma}
location of ((D+1)/2, (D+1)/2)
% in the original image.
%upsampling
upsampled image d7 = zeros(size of image); % It will simply upsample the image. ✓
Upsampling of the downsampled
% image to have exactly the size of the original image by filling zeros in
% the missing pixels.
p = 1; % initializing the value of p.
for i = (downsampling factor+1)/2:downsampling factor:size of image % for loop as per ✓
the given condition.
    q = 1; % initializing the value of q.
    for j = (downsampling factor+1)/2:downsampling factor:size of image
    upsampled image d7(i,j) = downsampled image <math>d7(p,q); % for loop as per the given \checkmark
condition.
        q = q + 1; % incrementing the value of q.
    end % end for loop.
    p = p + 1; % incrementing the value of q.
end % end for loop.
% Convolution with block shape for D=7
block filter d7 = [1 1 1 1 1 1]; % defining the filter as per the given problem.
block convoluted image d7 = upsampled image d7; % the block convolutional filter shold \checkmark
be of the same size as
```

```
% of the upsampled image.
for i = 1:size of image % for loop condition
block convoluted image d7(i,:) = conv(block convoluted image <math>d7(i,:), block filter d7, \checkmark
'same'); % Here, the
    % convolutions of the rows of the matrix. convolution operation for rows.
end % for loop ends
for j = 1:size of image % for loop condition
block convoluted image d7(:,j) = conv(block convoluted image <math>d7(:,j), \checkmark
block filter d7.', 'same'); % Here, the
    % convolutions of the rows of the matrix. convolution operation for columns.
end % for loop ends
% Convolution with triangle shape for D=7
triangle filter d7 = (1/7) * [1 2 3 4 5 6 7 5 4 3 2 1]; % defining the filter as per \checkmark
the given problem.
triangle convoluted image d7 = upsampled image d7; % the block convolutional filter \checkmark
shold be of the same size
% as of the upsampled image.
for i = 1:size of image % for loop condition
    triangle convoluted image d7(i,:) = conv(triangle convoluted image <math>d7(i,:), \checkmark
triangle filter d7, 'same');
    % Here, the
    % convolutions of the rows of the matrix. convolution operation for rows.
end % for loop ends
for j = 1:size of image % for loop condition
    triangle convoluted image d7(:,j) = conv(triangle convoluted image <math>d7(:,j), \checkmark
triangle filter d7.', 'same');
    % Here, the
    % convolutions of the rows of the matrix. convolution operation for columns.
end % for loop ends
% Figure plotting for D=7
figure; % figure creates figure graphics objects. figure objects are the individual ✓
windows on the screen
% in which MATLAB displays graphical output.
subplot(2,2,1); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=2, p=1.
imshow(input image); % It will display the gray-scale image in the figure.
title('Original image'); % It will add the specified title for the current plot.
subplot(2,2,2); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark
creates axes in the
% position specified by p. Here, m=n=p=2.
imshow(uint8(upsampled image d7)); % It will display the gray-scale image in the figure \checkmark
and
% it will convert each and every pixel value of the input image into the range of 0 to \checkmark
title('Upsampled image (D=7)'); % It will add the specified title for the current plot.
```

- $\operatorname{subplot}(2,2,3)$; % $\operatorname{subplot}(m,\ n,\ p)$ divides the current figure into an m-by-n grid and \mathbf{k}' creates axes in the
- % position specified by p. Here, m=n=2, p=3.
- imshow(uint8(block_convoluted_image_d7)); % It will display the gray-scale image in the

 figure and
- \$ it will convert each and every pixel value of the input image into the range of 0 to $\rlap/$ 255.
- title('Interpolated image using Nearest Neighbour (D=7)'); % It will add the specified \checkmark title for the current plot.
- subplot(2,2,4); % subplot(m, n, p) divides the current figure into an m-by-n grid and \checkmark creates axes in the
- % position specified by p. Here, m=n=2, p=4.
- imshow(uint8(triangle_convoluted_image_d7)); % It will display the gray-scale image in \checkmark the figure and
- % it will convert each and every pixel value of the input image into the range of 0 to \checkmark 255.
- title('Image using Bilinear interpolation (D=7)'); % It will add the specified title \checkmark for the current plot.

Figure 1:

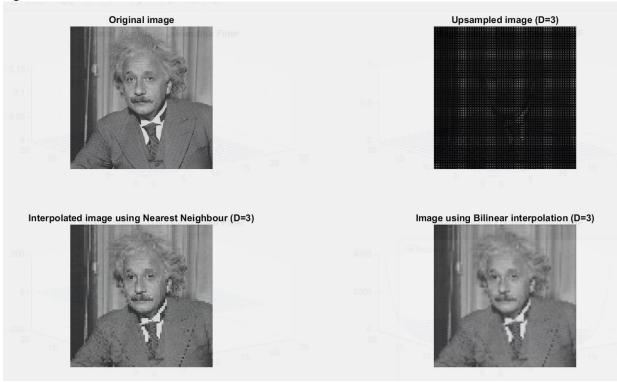
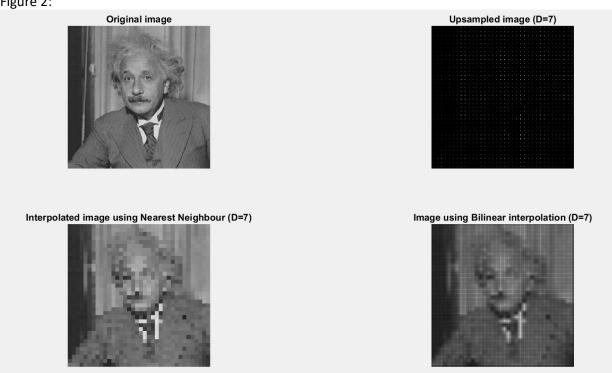


Figure 2:



Q5 Observation:

After downsampling the original image, we determined that the Nearest Neighbor Interpolation strategy was the most effective in restoring the image to its original condition. The bilinear interpolation is followed by the up sampled image. When applied to the original image, the Nearest Neighbor algorithm creates blocky results. This is because in the closest neighbor method, we find the next-closest pixel and assume its intensity value. As a result, the image takes on a blocky appearance. When compared to closest neighbor interpolation, bilinear interpolation estimates a new pixel value using a distance weighted average of the D - nearest pixel values, resulting in a smoother image.