

**Lab Report 4 : Image Filtering (Mean Filter, Median Filter  
and Gaussian Filter),  
Image Resizing (Replication Method, Linear Interpolation  
Method and Pixel Mapping Method)**

**Course Title:** Digital Image Processing Lab

**Course Code:** CSE-406



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## Experiment Name

- a) Illustration of image filtering using Mean Filter method, Median Filter method, and Gaussian Filter method.
- b) Image Resizing using Replication method, Linear Interpolation method, and Pixel Mapping method.

## Objectives

- a) To apply and compare different image filtering methods (Mean Filter, Median Filter, Gaussian Filter) on a given image and analyze their effects.
- b) To explore and implement various image resizing methods (Replication, Linear Interpolation, Pixel Mapping) and evaluate their performance on image scaling.

### Experiment 1: Image filtering using Mean Filter, Median Filter, and Gaussian Filter method

#### MATLAB Code:

Listing 1: Image\_Filtering.m

```

1 % Load the image
2 image = imread('image.jpg');
3 image_gray = rgb2gray(image); % Convert to grayscale if needed
4
5 % Define the mean filter (5x5 kernel)
6 mean_filter = fspecial('average', [5 5]);
7
8 % Apply the mean filter
9 mean_filtered = imfilter(image_gray, mean_filter);
10
11 % Apply the median filter (5x5 kernel)
12 median_filtered = medfilt2(image_gray, [5 5]);
13
14 % Define the Gaussian filter (5x5 kernel, sigma=1)
15 gaussian_filter = fspecial('gaussian', [5 5], 1);
16
17 % Apply the Gaussian filter
18 gaussian_filtered = imfilter(image_gray, gaussian_filter);
19
20 % Plot all the images in one frame
21 figure;
22
23 % Original image
24 subplot(2, 2, 1);
25 imshow(image_gray);
26 title('Original Image');

```

```

27 % Mean filtered image
28 subplot(2, 2, 2);
29 imshow(mean_filtered);
30 title('Mean Filtered Image');
31
32 % Median filtered image
33 subplot(2, 2, 3);
34 imshow(median_filtered);
35 title('Median Filtered Image');
36
37 % Gaussian filtered image
38 subplot(2, 2, 4);
39 imshow(gaussian_filtered);
40 title('Gaussian Filtered Image');
41
42 % Save the composite figure
43 saveas(gcf, 'filtered_images_composite.jpg');

```

## Python Code:

Listing 2: Image\_Filtering.py

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 from PIL import Image
4 from scipy.ndimage import convolve, median_filter,
5     gaussian_filter
6 import os
7
8 def mean_filter(image_array, kernel_size):
9     if kernel_size % 2 == 0:
10         raise ValueError("Kernel size must be odd.")
11
12     kernel = np.ones((kernel_size, kernel_size),
13                      dtype=np.float32) / (kernel_size * kernel_size)
14     filtered_image_array = np.zeros_like(image_array)
15
16     for i in range(3): # Assuming image_array has 3 channels
17         (RGB)
18         filtered_image_array[:, :, i] = convolve(image_array[:, :, i], kernel, mode='reflect')
19
20     return filtered_image_array
21
22 def apply_median_filter(image_array, kernel_size):
23     if kernel_size % 2 == 0:
24         raise ValueError("Kernel size must be odd.")
25
26     filtered_image_array = np.zeros_like(image_array)

```

```

25     for i in range(3): # Assuming image_array has 3 channels
26         # (RGB)
27         filtered_image_array[:, :, i] =
28             median_filter(image_array[:, :, i], size=kernel_size)
29
30     return filtered_image_array
31
32
33 def apply_gaussian_filter(image_array, sigma):
34     filtered_image_array = np.zeros_like(image_array)
35
36     for i in range(3): # Assuming image_array has 3 channels
37         # (RGB)
38         filtered_image_array[:, :, i] =
39             gaussian_filter(image_array[:, :, i], sigma=sigma)
40
41     return filtered_image_array
42
43
44 # Load an image using PIL
45 image_path = 'image.jpg' # Replace with your image path
46 image = Image.open(image_path)
47
48 # Convert the image to a NumPy array
49 image_array = np.array(image)
50
51
52 # Ensure the image is RGB
53 if image_array.ndim == 3 and image_array.shape[2] == 3:
54     # Apply filters
55     mean_filtered_array = mean_filter(image_array, kernel_size=5)
56     median_filtered_array = apply_median_filter(image_array,
57           kernel_size=5)
58     gaussian_filtered_array = apply_gaussian_filter(image_array,
59           sigma=1.0)
60
61 else:
62     raise ValueError("The input image is not an RGB image.")
63
64 # Create outputs directory in the same directory as the script
65 output_dir = os.path.join(os.getcwd(), 'outputs')
66 if not os.path.exists(output_dir):
67     os.makedirs(output_dir)
68
69 # Convert results back to images
70 mean_filtered_image =
71     Image.fromarray(np.uint8(mean_filtered_array))
72 median_filtered_image =
73     Image.fromarray(np.uint8(median_filtered_array))
74 gaussian_filtered_image =
75     Image.fromarray(np.uint8(gaussian_filtered_array))
76
77 # Save the results
78 mean_filtered_image.save(os.path.join(output_dir,
79     'mean_filtered_image.jpg'))

```

```
66 median_filtered_image.save(os.path.join(output_dir,
67     'median_filtered_image.jpg'))
68 gaussian_filtered_image.save(os.path.join(output_dir,
69     'gaussian_filtered_image.jpg'))
70
71 # Display the images
72 plt.figure(figsize=(15, 10))
73
74 plt.subplot(2, 2, 1)
75 plt.title('Original Image')
76 plt.imshow(image_array)
77 plt.axis('off')
78
79 plt.subplot(2, 2, 2)
80 plt.title('Mean Filtered Image')
81 plt.imshow(mean_filtered_array)
82 plt.axis('off')
83
84 plt.subplot(2, 2, 3)
85 plt.title('Median Filtered Image')
86 plt.imshow(median_filtered_array)
87 plt.axis('off')
88
89 plt.subplot(2, 2, 4)
90 plt.title('Gaussian Filtered Image')
91 plt.imshow(gaussian_filtered_array)
92 plt.axis('off')
93
94 plt.show()
```

## Output:

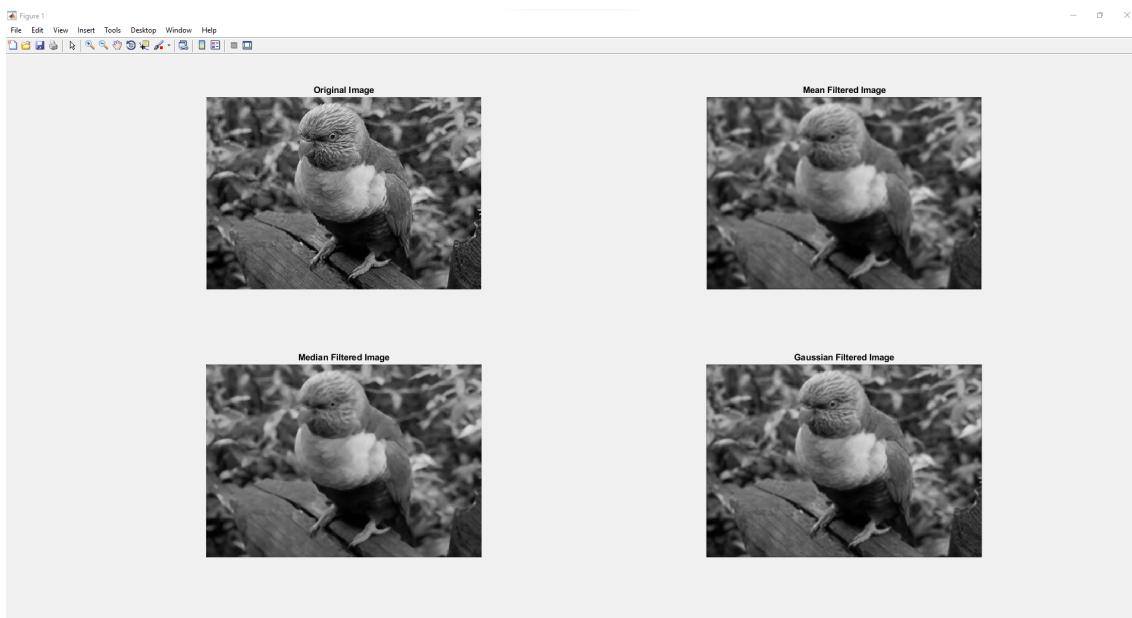


Figure 1: Image Filtering of Gray Scale Image



Figure 2: Image Filtering of RGB Image

## Experiment 2: Image Resizing using Replication, Linear Interpolation and Pixel Mapping method

### MATLAB Code:

Listing 3: Image\_Resizing.m

```

1 % Step 1: Get User Input for Image File and New Size
2
3 % Prompt user for the image file name
4 imageFileName = input('Enter the image file name (e.g.,
5   'peppers.png'): ', 's');
6
7 % Read the image
8 originalImage = imread(imageFileName);
9
10 % Prompt user for the new size
11 newHeight = input('Enter the new height for resizing: ');
12 newWidth = input('Enter the new width for resizing: ');
13 newSize = [newHeight, newWidth]; % New size [height, width]
14
15 % Convert to double if needed for better interpolation results
16 if ~isa(originalImage, 'double')
17     originalImage = im2double(originalImage);
18 end
19
20 % Step 2: Resize using Replication Method (Nearest-Neighbor
21   Interpolation)
22 resizedImageReplication = imresize(originalImage, newSize,
23   'nearest');
24
25 % Step 3: Resize using Linear Interpolation Method (Bilinear
26   Interpolation)
27 resizedImageLinear = imresize(originalImage, newSize,
28   'bilinear');
29
30 % Step 4: Resize using Pixel Skipping Method
31 % Pixel skipping involves subsampling
32
33 % Calculate the skipping ratio
34 originalSize = size(originalImage);
35 rowRatio = originalSize(1) / newSize(1);
36 colRatio = originalSize(2) / newSize(2);
37
38 % Use floor to ensure indices are within bounds
39 rowSkip = floor(rowRatio);
40 colSkip = floor(colRatio);
41
42 % Generate the resized image by skipping pixels
43 resizedImagePixelSkipping = originalImage(1:rowSkip:end,
44   1:colSkip:end, :);

```

```

39 % Adjust size to match exactly if necessary
40 % Resize pixel skipping result to newSize using imresize if
41 % necessary
42 resizedImagePixelSkipping = imresize(resizedImagePixelSkipping,
43     newSize, 'nearest');
44
45 % Step 5: Display the Results
46 figure;
47
48 % Display original image
49 subplot(2, 2, 1);
50 imshow(originalImage);
51 title('Original Image');
52
53 % Display resized image using replication (nearest neighbor)
54 subplot(2, 2, 2);
55 imshow(resizedImageReplication);
56 title('Resized Image (Replication)');
57
58 % Display resized image using linear interpolation
59 subplot(2, 2, 3);
60 imshow(resizedImageLinear);
61 title('Resized Image (Linear Interpolation)');
62
63 % Display resized image using pixel skipping method
64 subplot(2, 2, 4);
65 imshow(resizedImagePixelSkipping);
66 title('Resized Image (Pixel Skipping)');

```

## Python Code:

Listing 4: Image\_Resizing.py

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 from PIL import Image
4
5 def linear_interpolation_resize(image_array, new_width,
6     new_height):
7     """
8         Resize an image using linear interpolation method.
9     """
10    old_height, old_width, channels = image_array.shape
11    resized_image_array = np.zeros((new_height, new_width,
12        channels), dtype=image_array.dtype)
13    row_scale = old_height / new_height
14    col_scale = old_width / new_width
15
16    for y in range(new_height):
17        for x in range(new_width):

```

```

16         orig_y = y * row_scale
17         orig_x = x * col_scale
18         y0 = int(orig_y)
19         x0 = int(orig_x)
20         y1 = min(y0 + 1, old_height - 1)
21         x1 = min(x0 + 1, old_width - 1)
22         dy = orig_y - y0
23         dx = orig_x - x0
24         w00 = (1 - dx) * (1 - dy)
25         w01 = dx * (1 - dy)
26         w10 = (1 - dx) * dy
27         w11 = dx * dy
28
29     for c in range(channels):
30         pixel_value = (w00 * image_array[y0, x0, c] +
31                         w01 * image_array[y0, x1, c] +
32                         w10 * image_array[y1, x0, c] +
33                         w11 * image_array[y1, x1, c])
34         resized_image_array[y, x, c] = pixel_value
35     return resized_image_array
36
37 def pixel_skipping_resize(image_array, new_width, new_height):
38     """
39     Resize an image using the pixel skipping method.
40     """
41     old_height, old_width, channels = image_array.shape
42     resized_image_array = np.zeros((new_height, new_width,
43                                    channels), dtype=image_array.dtype)
44     row_scale = old_height / new_height
45     col_scale = old_width / new_width
46
47     for y in range(new_height):
48         for x in range(new_width):
49             src_y = int(y * row_scale)
50             src_x = int(x * col_scale)
51             resized_image_array[y, x] = image_array[src_y, src_x]
52     return resized_image_array
53
54 def replicate_resize(image_array, new_width, new_height):
55     """
56     Resize an image using the replication method.
57     """
58     old_height, old_width, channels = image_array.shape
59     resized_image_array = np.zeros((new_height, new_width,
60                                    channels), dtype=image_array.dtype)
61     row_scale = old_height / new_height
62     col_scale = old_width / new_width
63
64     for y in range(new_height):
65         for x in range(new_width):
66             src_y = int(y * row_scale)

```

```

65         src_x = int(x * col_scale)
66         # Use nearest neighbor replication
67         resized_image_array[y, x] = image_array[src_y, src_x]
68     return resized_image_array
69
70 # Load an image using PIL
71 image_path = 'rgb-image.jpg' # Replace with your image path
72 try:
73     image = Image.open(image_path)
74 except FileNotFoundError:
75     print(f"Error: The file {image_path} was not found.")
76     exit()
77
78 image_array = np.array(image)
79
80 # Get and print the original image dimensions
81 original_height, original_width, _ = image_array.shape
82 print(f"Original image dimensions:
83     {original_width}x{original_height}")
84
85 # Input the new dimensions for resizing
86 try:
87     new_width = int(input("Enter the new width: "))
88     new_height = int(input("Enter the new height: "))
89 except ValueError:
90     print("Error: Please enter valid integer values for width
91         and height.")
92     exit()
93
94 # Perform the resizing using different methods
95 resized_image_linear = linear_interpolation_resize(image_array,
96             new_width, new_height)
97 resized_image_pixel_skip = pixel_skipping_resize(image_array,
98             new_width, new_height)
99 resized_image_replicate = replicate_resize(image_array,
100             new_width, new_height)
101
102 # Convert back to image for saving purposes
103 Image.fromarray(np.uint8(resized_image_linear)).
104     save('outputs/linear_interpolation_resized.jpg')
105 Image.fromarray(np.uint8(resized_image_pixel_skip)).
106     save('outputs/pixel_skipping_resized.jpg')
107 Image.fromarray(np.uint8(resized_image_replicate)).
108     save('outputs/replication_resized.jpg')
109
110 # Plot and compare results
111 plt.figure(figsize=(15, 10))
112
113 # Original Image
114 plt.subplot(2, 2, 1)
115 plt.title('Original Image')

```

```
108 plt.imshow(image_array)
109 plt.axis('off')
110 plt.text(0.5, -0.1, f"Dimensions:
111     {original_width}x{original_height}", ha='center',
112         va='center', fontsize=12, transform=plt.gca().transAxes)
113
114 # Resized using Linear Interpolation
115 plt.subplot(2, 2, 2)
116 plt.title('Resized Image using Linear Interpolation')
117 plt.imshow(resized_image_linear.astype(np.uint8))
118 plt.axis('off')
119 plt.text(0.5, -0.1, f"Dimensions: {new_width}x{new_height}",
120         ha='center', va='center', fontsize=12,
121             transform=plt.gca().transAxes)
122
123 # Resized using Pixel Skipping
124 plt.subplot(2, 2, 3)
125 plt.title('Resized Image using Pixel Skipping')
126 plt.imshow(resized_image_pixel_skip.astype(np.uint8))
127 plt.axis('off')
128 plt.text(0.5, -0.1, f"Dimensions: {new_width}x{new_height}",
129         ha='center', va='center', fontsize=12,
130             transform=plt.gca().transAxes)
131
132 # Resized using Replication
133 plt.subplot(2, 2, 4)
134 plt.title('Resized Image using Replication')
135 plt.imshow(resized_image_replicate.astype(np.uint8))
136 plt.axis('off')
137 plt.text(0.5, -0.1, f"Dimensions: {new_width}x{new_height}",
138         ha='center', va='center', fontsize=12,
139             transform=plt.gca().transAxes)
140
141 plt.tight_layout()
142 plt.show()
```

## Output:

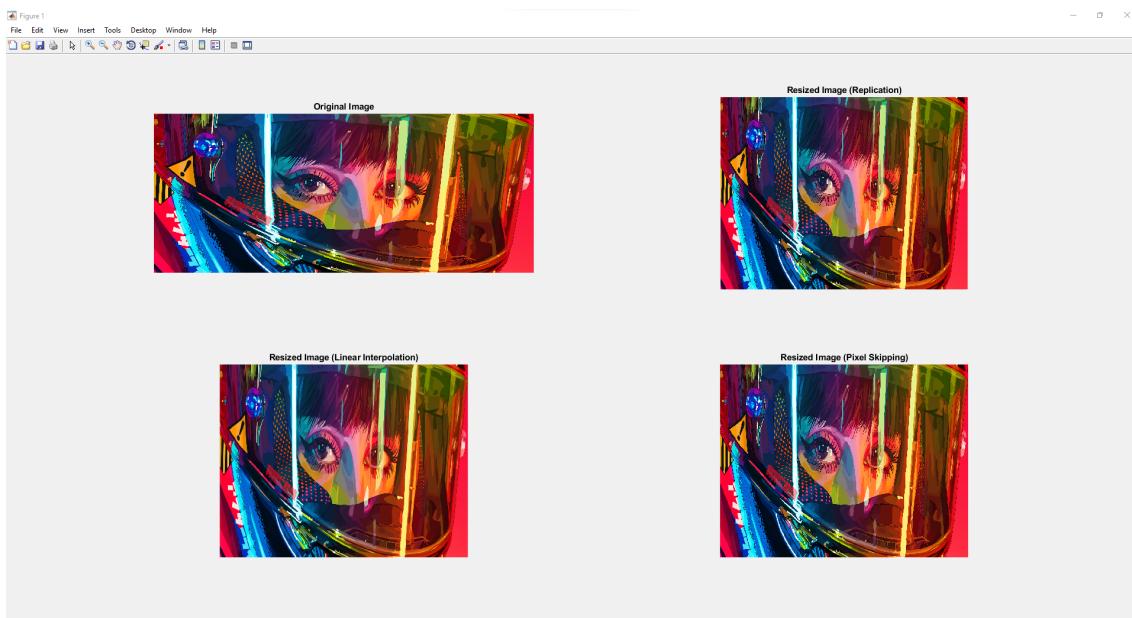


Figure 3: Image Resizing Output for Matlab Code

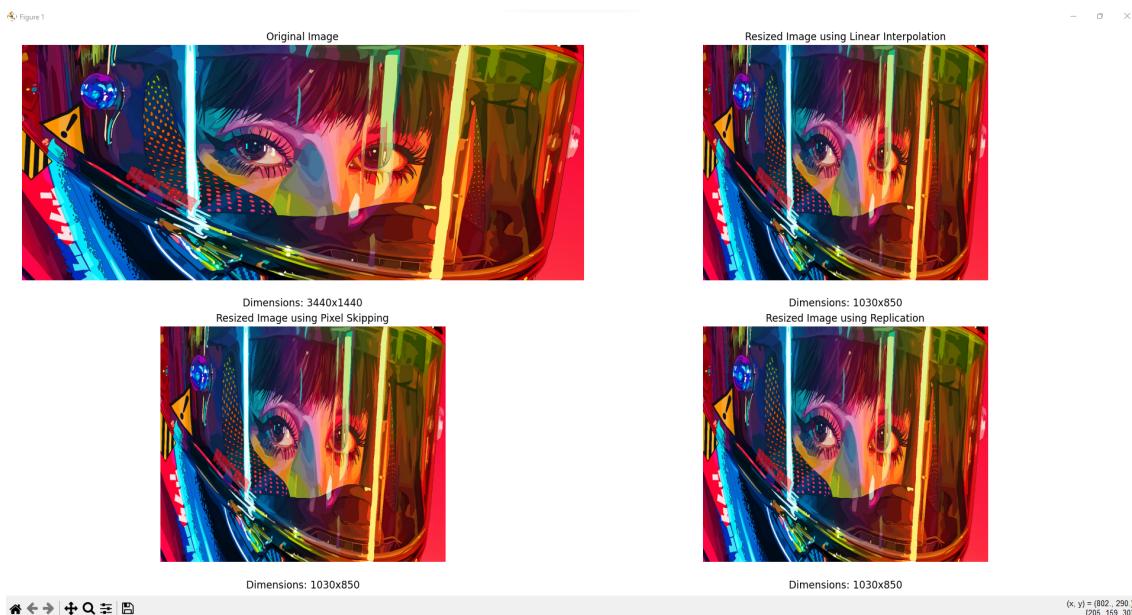


Figure 4: Image Resizing Output for Python Code

## Results

### A) Image Filtering Results

- **Original Image** – The original grayscale and RGB both images are used as the base for filtering.
- **Mean Filtered Image** – After applying the Mean Filter, we observe that noise in the image is smoothed, but fine details may also be slightly blurred.
- **Median Filtered Image** – The Median Filter is effective in removing salt-and-pepper noise while preserving the edges of the image.
- **Gaussian Filtered Image** – Gaussian Filter smooths the image by reducing high-frequency components while introducing a Gaussian blur effect, depending on the sigma value used.

**Comparison:** Each filter shows different levels of smoothing and noise reduction, with the Median Filter being most effective for removing noise without sacrificing details, and Gaussian Filter introducing smoothness while preserving more of the original image's structure compared to the Mean Filter.

### B) Image Resizing Results

- **Replication Method** – This method results in a pixelated or blocky appearance, as it simply replicates the nearest pixel value for scaling.
- **Linear Interpolation Method** – Produces a smoother resized image by calculating weighted averages of neighboring pixel values, reducing the pixelation effect.
- **Pixel Mapping Method** – Involves subsampling pixels, which can lead to loss of detail but ensures fast and efficient resizing for certain applications.

**Comparison:** Linear interpolation provides the best quality among the methods, preserving more detail and producing smooth transitions in pixel values. Replication can be fast but results in blockiness, while pixel mapping is simple but may sacrifice accuracy.