

Part 1

```
import cv2
import numpy as np
import matplotlib.pyplot as plt

img = cv2.imread(r"D:\MIT WPU B.Tech Data\3nd Year Data\IPPR\low_contrast_image.png", cv2.IMREAD_GRAYSCALE)

hist_original = cv2.calcHist(
    [img],
    [0],
    None,
    [256],
    [0, 256]
)

equalized_img = cv2.equalizeHist(img)

hist_equalized = cv2.calcHist(
    [equalized_img],
    [0],
    None,
    [256],
    [0, 256]
)

plt.figure(figsize=(10, 8))

plt.subplot(2, 2, 1)
plt.title("Original Grayscale Image")
plt.imshow(img, cmap='gray')
plt.axis("off")

plt.subplot(2, 2, 2)
plt.title("Equalized Image")
plt.imshow(equalized_img, cmap='gray')
plt.axis("off")

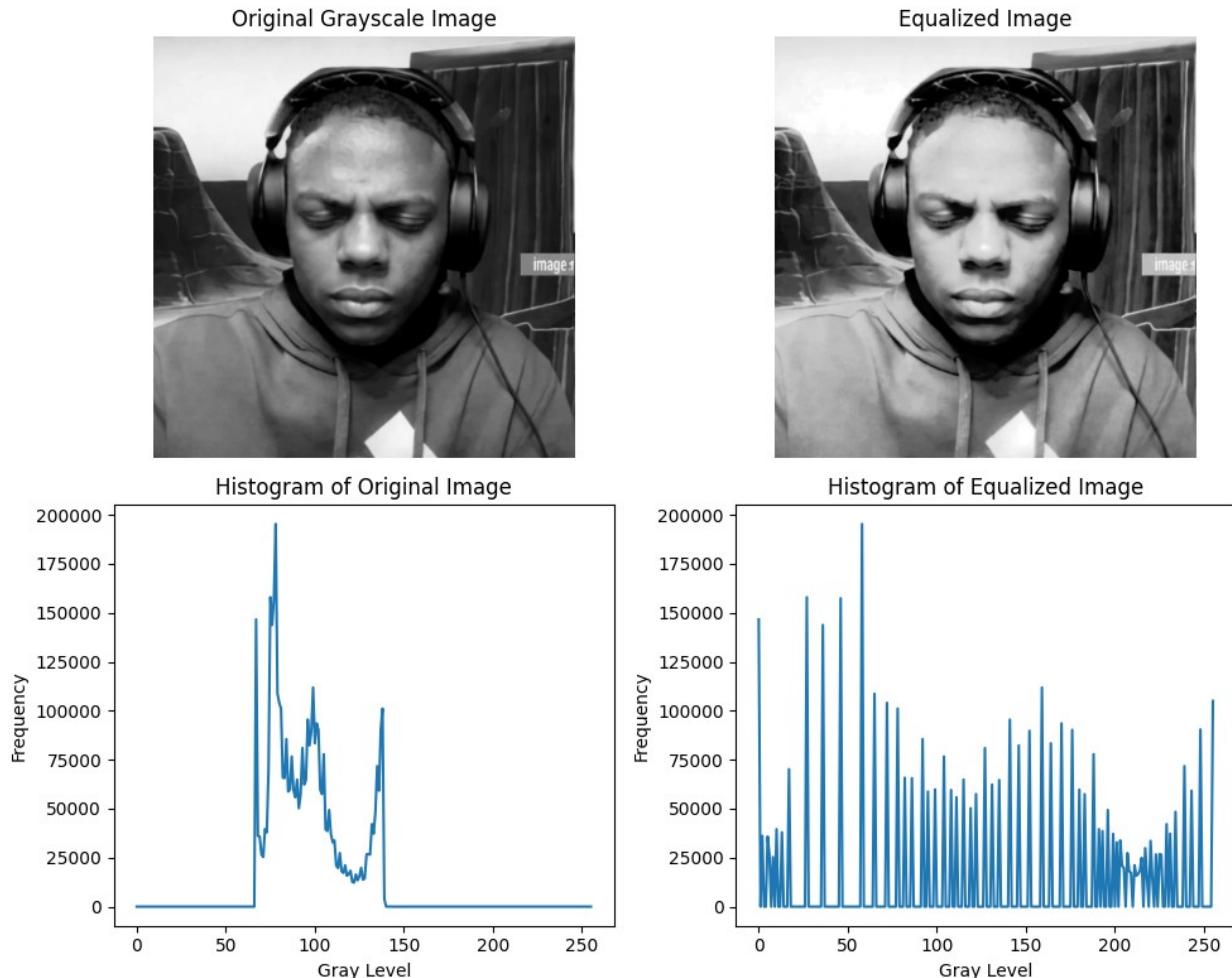
plt.subplot(2, 2, 3)
plt.title("Histogram of Original Image")
plt.plot(hist_original)
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.subplot(2, 2, 4)
plt.title("Histogram of Equalized Image")
plt.plot(hist_equalized)
plt.xlabel("Gray Level")
```

```

plt.ylabel("Frequency")
plt.tight_layout()
plt.show()

```



Part 2

```

img_rgb = cv2.imread(r"D:\MIT WPU B.Tech Data\3nd Year Data\IPPR\charlie-kirk-charlie-kirk-meme.gif")
img_rgb = cv2.cvtColor(img_rgb, cv2.COLOR_BGR2RGB)

R, G, B = cv2.split(img_rgb)

hist_R = cv2.calcHist([R], [0], None, [256], [0,256])
hist_G = cv2.calcHist([G], [0], None, [256], [0,256])
hist_B = cv2.calcHist([B], [0], None, [256], [0,256])

```

```

R_eq = cv2.equalizeHist(R)
G_eq = cv2.equalizeHist(G)
B_eq = cv2.equalizeHist(B)

img_rgb_eq = cv2.merge((R_eq, G_eq, B_eq))

hist_R_eq = cv2.calcHist([R_eq], [0], None, [256], [0,256])
hist_G_eq = cv2.calcHist([G_eq], [0], None, [256], [0,256])
hist_B_eq = cv2.calcHist([B_eq], [0], None, [256], [0,256])

plt.figure(figsize=(8,4))
plt.subplot(1,2,1)
plt.title("Original RGB")
plt.imshow(img_rgb)
plt.axis("off")

plt.subplot(1,2,2)
plt.title("Equalized RGB")
plt.imshow(img_rgb_eq)
plt.axis("off")
plt.show()

plt.figure(figsize=(12, 8))

plt.subplot(3, 2, 1)
plt.title("Original Red Channel")
plt.plot(hist_R, color='r')
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.subplot(3, 2, 2)
plt.title("Equalized Red Channel")
plt.plot(hist_R_eq, color='r')
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.subplot(3, 2, 3)
plt.title("Original Green Channel")
plt.plot(hist_G, color='g')
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.subplot(3, 2, 4)
plt.title("Equalized Green Channel")
plt.plot(hist_G_eq, color='g')
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.subplot(3, 2, 5)

```

```
plt.title("Original Blue Channel")
plt.plot(hist_B, color='b')
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

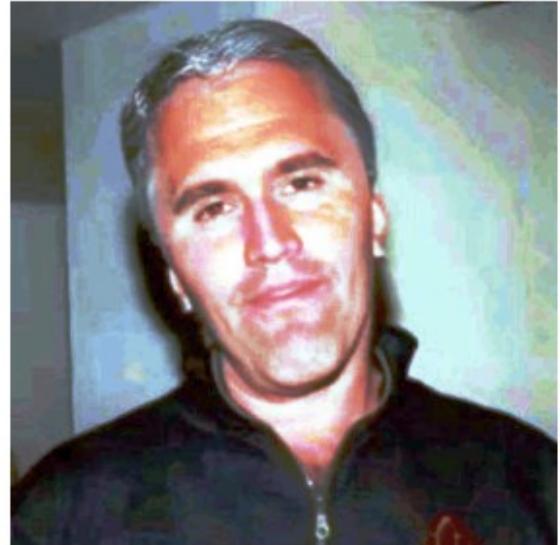
plt.subplot(3, 2, 6)
plt.title("Equalized Blue Channel")
plt.plot(hist_B_eq, color='b')
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

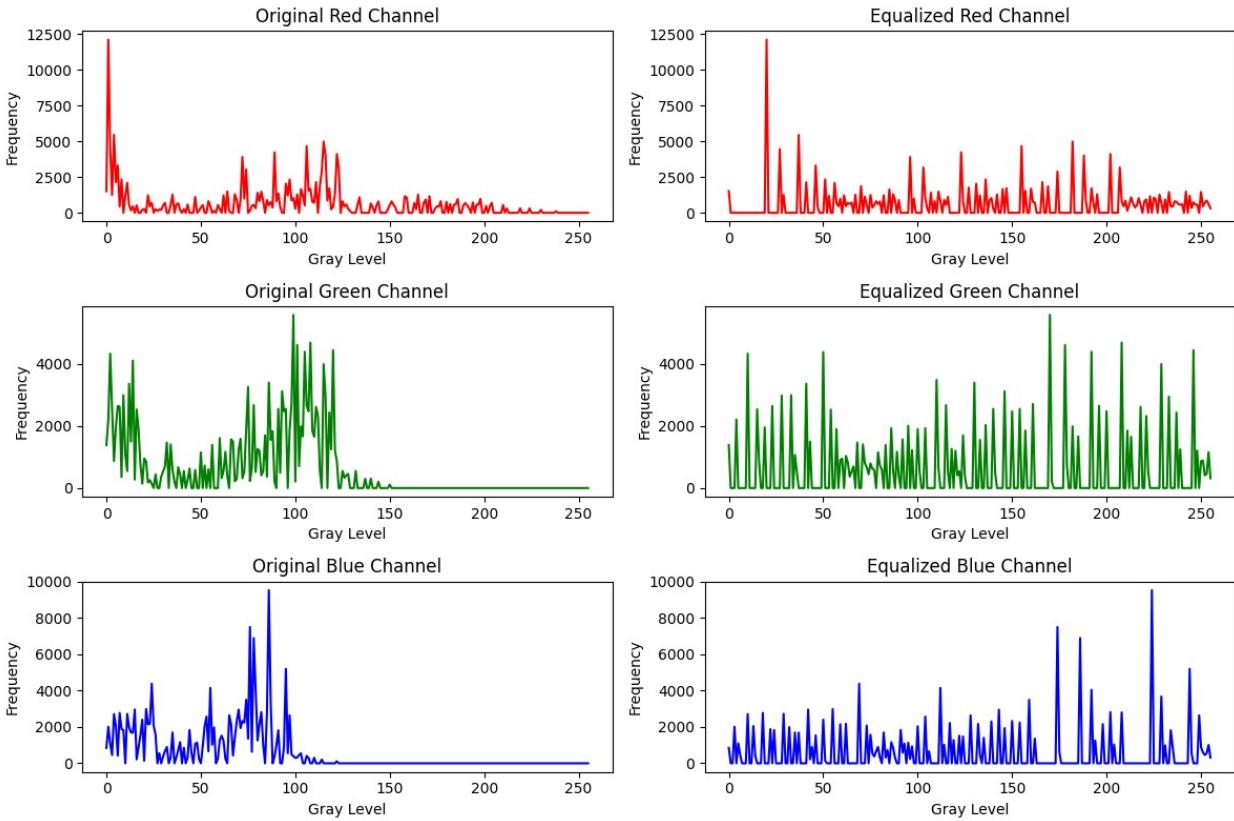
plt.tight_layout()
plt.show()
```

Original RGB



Equalized RGB





Convert the RGB image to HSI

```
img_hsv = cv2.cvtColor(img_rgb, cv2.COLOR_RGB2HSV)
H, S, V = cv2.split(img_hsv)

hist_V = cv2.calcHist([V], [0], None, [256], [0,256])
V_eq = cv2.equalizeHist(V)

hist_V_eq = cv2.calcHist([V_eq], [0], None, [256], [0,256])
img_hsv_eq = cv2.merge((H, S, V_eq))
```

Convert the HSI image back to RGB

```
img_rgb_hsi_eq = cv2.cvtColor(img_hsv_eq, cv2.COLOR_HSV2RGB)

plt.figure(figsize=(8,4))
plt.subplot(1,2,1)
plt.title("Original RGB")
plt.imshow(img_rgb)
plt.axis("off")

plt.subplot(1,2,2)
plt.title("HSI (I Equalized)")
```

```
plt.imshow(img_rgb_hsi_eq)
plt.axis("off")
plt.show()

plt.figure(figsize=(8,4))
plt.subplot(1,2,1)
plt.title("Original 'I' Component (HSI image)")
plt.plot(hist_V)

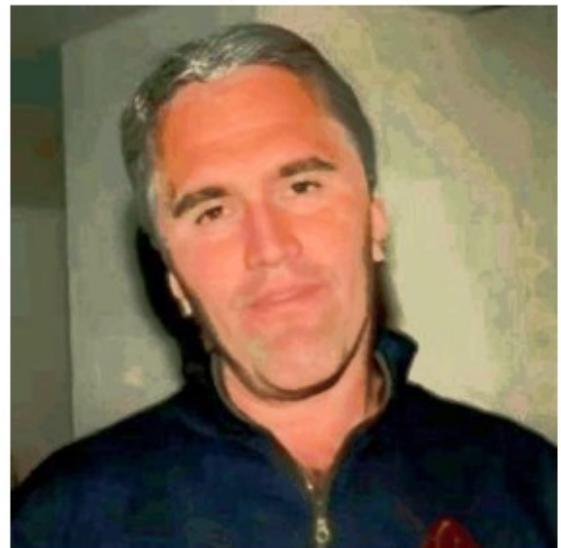
plt.subplot(1,2,2)
plt.title("Equalized 'I' Component (HSI image)")
plt.plot(hist_V_eq)

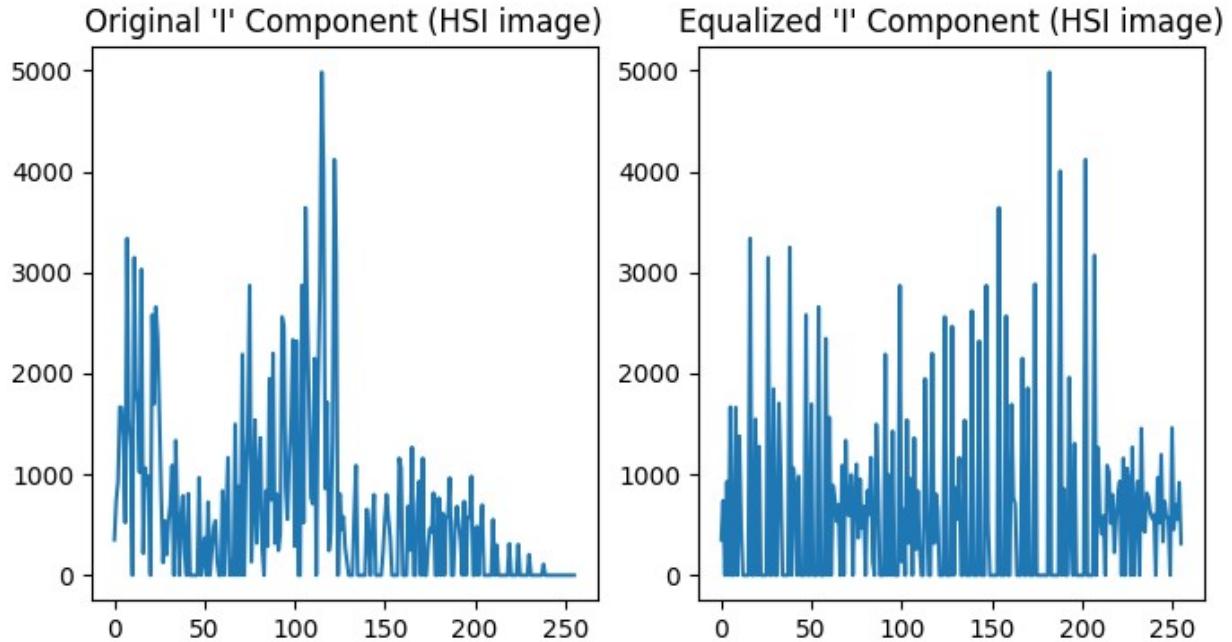
plt.show()
```

Original RGB



HSI (I Equalized)





Calculating the histogram and equalizing the image manually

```



```

```
hist_eq[equalized_img[i, j]] += 1

plt.figure(figsize=(10, 8))

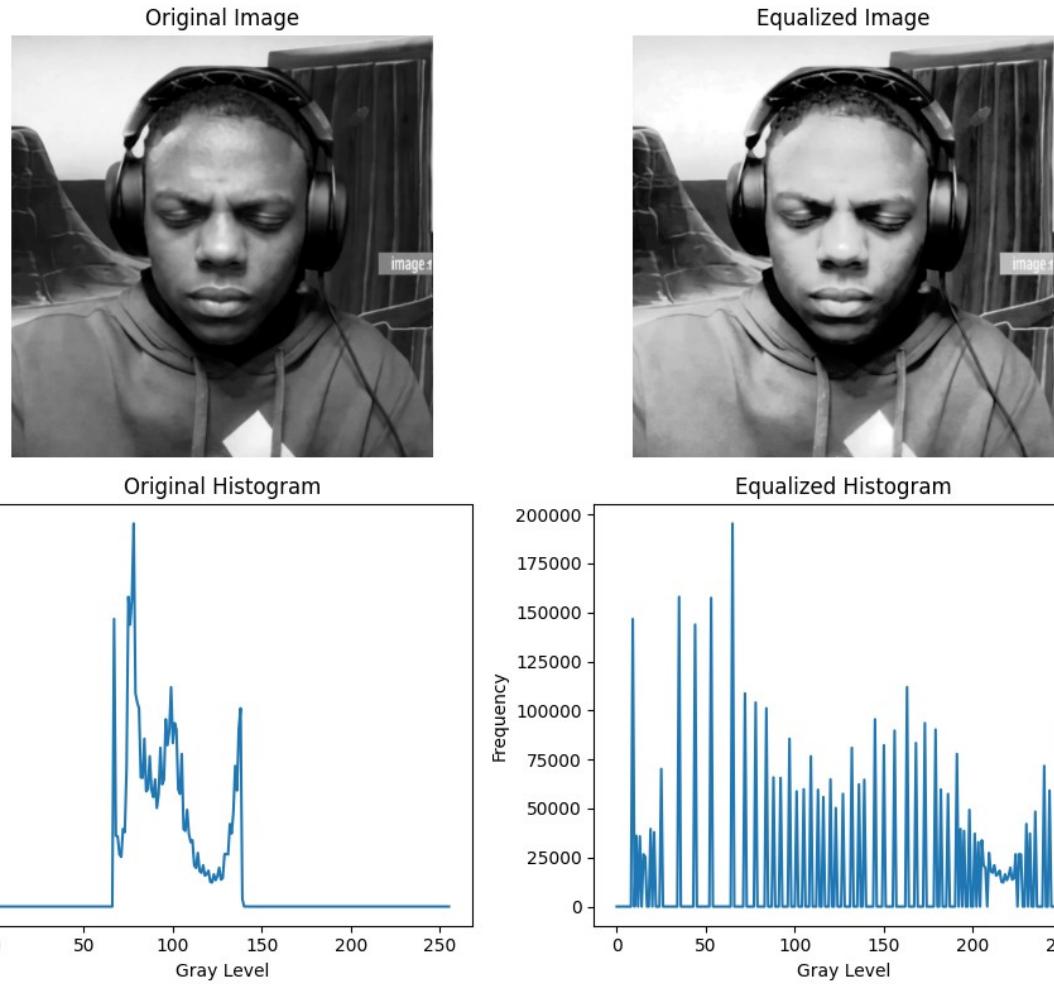
plt.subplot(2, 2, 1)
plt.title("Original Image")
plt.imshow(img, cmap="gray")
plt.axis("off")

plt.subplot(2, 2, 2)
plt.title("Equalized Image")
plt.imshow(equalized_img, cmap="gray")
plt.axis("off")

plt.subplot(2, 2, 3)
plt.title("Original Histogram")
plt.plot(hist)
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.subplot(2, 2, 4)
plt.title("Equalized Histogram")
plt.plot(hist_eq)
plt.xlabel("Gray Level")
plt.ylabel("Frequency")

plt.tight_layout()
plt.show()
```



Changing the bin component in the calcHist() function

```
hist_original = cv2.calcHist(
    [img],
    [0],
    None,
    [10],
    [0, 256]
)

bin_edges = np.linspace(0, 256, 11)

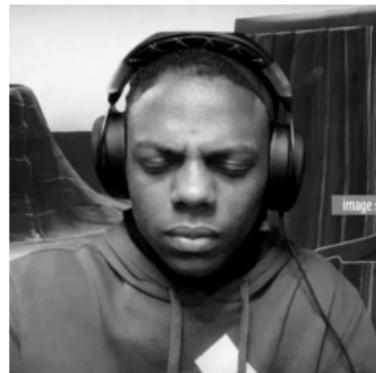
plt.figure(figsize=(8, 6))

plt.subplot(2, 1, 1)
plt.title("Original Grayscale Image")
plt.imshow(img, cmap='gray')
plt.axis("off")

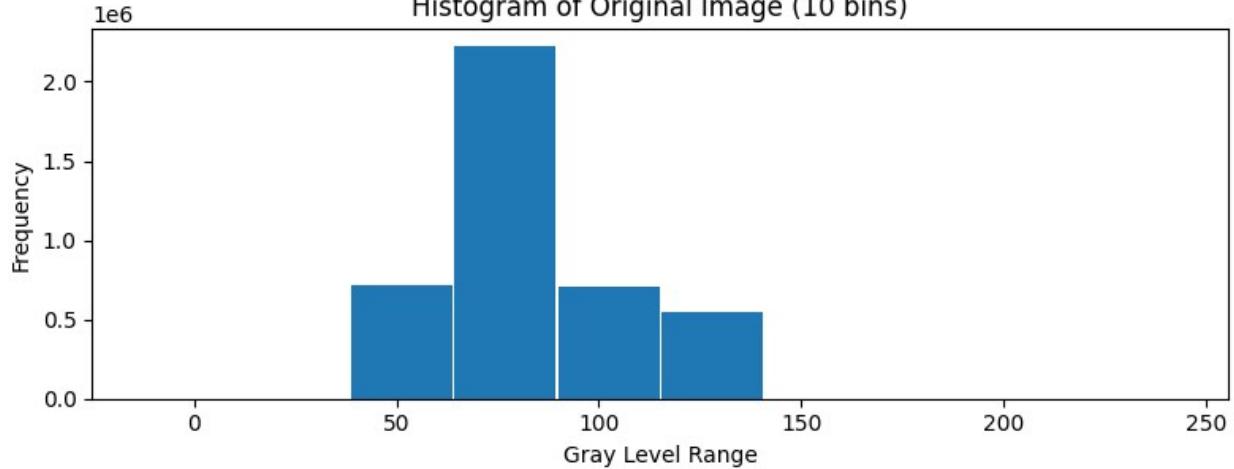
plt.subplot(2, 1, 2)
```

```
plt.title("Histogram of Original Image (10 bins)")  
plt.bar(bin_edges[:-1], hist_original.flatten(), width=25)  
plt.xlabel("Gray Level Range")  
plt.ylabel("Frequency")  
  
plt.tight_layout()  
plt.show()
```

Original Grayscale Image



Histogram of Original Image (10 bins)



```
In [21]: import cv2
import numpy as np
import matplotlib.pyplot as plt
```

Input Image

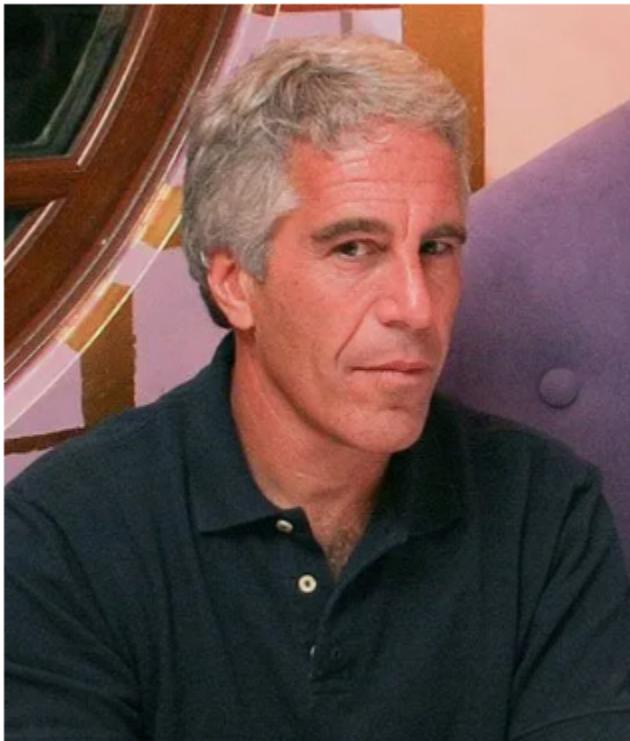
```
In [22]: img = cv2.imread(r"D:\MIT WPU B.Tech Data\3nd Year Data\IPPR\Jeffrey-Epstein-S

img_rgb = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)

plt.imshow(img_rgb)
plt.title("Original Image")
plt.axis("off")
```

Out[22]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))

Original Image



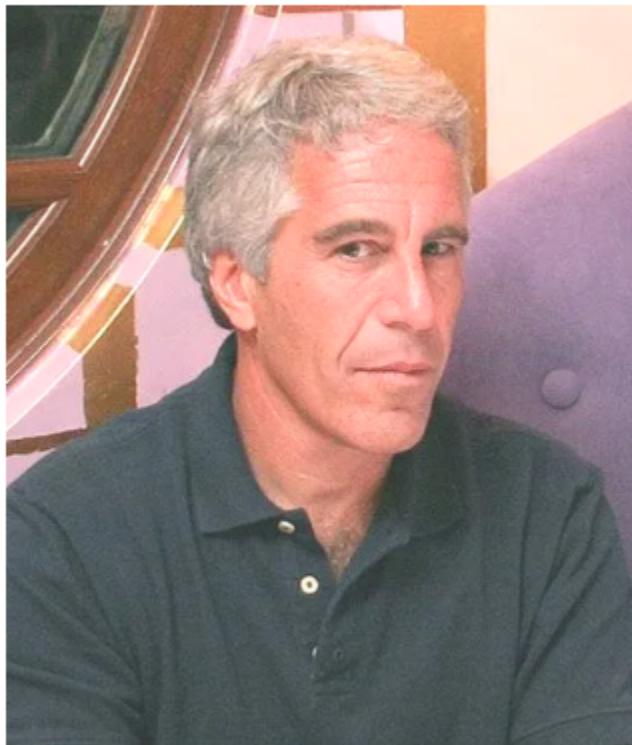
Arithmetic Operations

```
In [23]: bright = cv2.add(img, np.ones(img.shape, dtype='uint8') * 50)
bright_rgb = cv2.cvtColor(bright, cv2.COLOR_BGR2RGB)

plt.imshow(bright_rgb)
plt.title("Image Addition (Brightness Increased)")
plt.axis("off")
```

```
Out[23]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Image Addition (Brightness Increased)

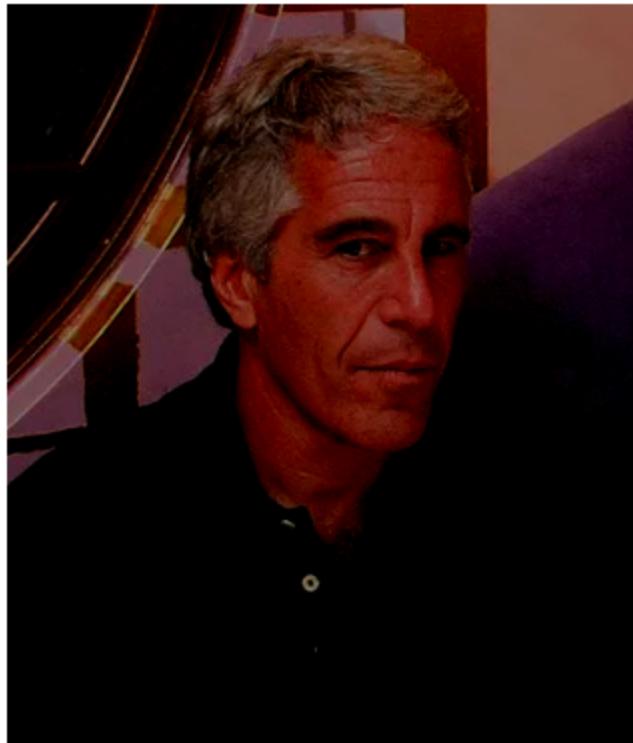


```
In [24]: dark = cv2.subtract(img, np.ones(img.shape, dtype='uint8') * 100)
dark_rgb = cv2.cvtColor(dark, cv2.COLOR_BGR2RGB)

plt.imshow(dark_rgb)
plt.title("Image Subtraction (Brightness Decreased)")
plt.axis("off")
```

```
Out[24]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Image Subtraction (Brightness Decreased)



```
In [25]: mul = cv2.multiply(img, np.array([1.2]))
mul_rgb = cv2.cvtColor(mul, cv2.COLOR_BGR2RGB)

plt.imshow(mul_rgb)
plt.title("Image Multiplication")
plt.axis("off")
```

```
Out[25]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Image Multiplication



```
In [26]: div = cv2.divide(img, np.array([1.5]))
div_rgb = cv2.cvtColor(div, cv2.COLOR_BGR2RGB)

plt.imshow(div_rgb)
plt.title("Image Division")
plt.axis("off")
```

```
Out[26]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Image Division

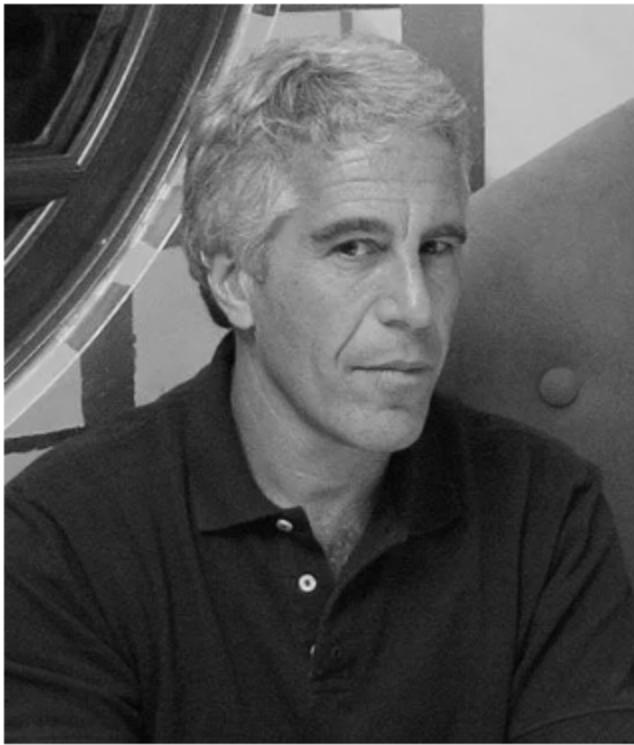


Logical Operations

```
In [27]: gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
plt.imshow(gray, cmap='gray')
plt.title("Grayscale Image")
plt.axis("off")
```

```
Out[27]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Grayscale Image



```
In [28]: _, mask = cv2.threshold(gray, 120, 255, cv2.THRESH_BINARY)
plt.imshow(mask, cmap='gray')
plt.title("Binary Mask")
plt.axis("off")
```

```
Out[28]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Binary Mask

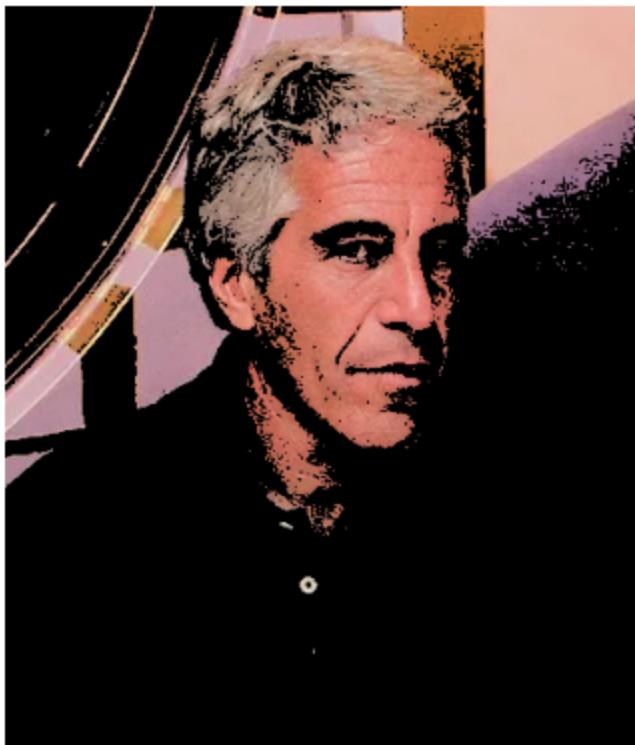


```
In [29]: bit_and = cv2.bitwise_and(img, img, mask=mask)
bit_and_rgb = cv2.cvtColor(bit_and, cv2.COLOR_BGR2RGB)

plt.imshow(bit_and_rgb)
plt.title("Bitwise AND")
plt.axis("off")
```

```
Out[29]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Bitwise AND

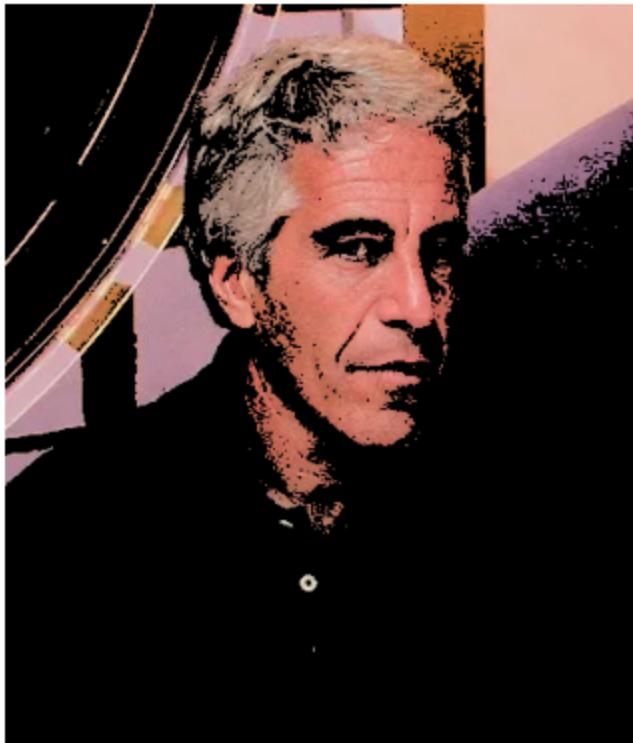


```
In [30]: bit_or = cv2.bitwise_or(img, img, mask=mask)
bit_or_rgb = cv2.cvtColor(bit_or, cv2.COLOR_BGR2RGB)

plt.imshow(bit_or_rgb)
plt.title("Bitwise OR")
plt.axis("off")
```

```
Out[30]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Bitwise OR



```
In [31]: bit_xor = cv2.bitwise_xor(img, img, mask=mask)
bit_xor_rgb = cv2.cvtColor(bit_xor, cv2.COLOR_BGR2RGB)

plt.imshow(bit_xor_rgb)
plt.title("Bitwise XOR")
plt.axis("off")
```

```
Out[31]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Bitwise XOR



```
In [32]: bit_not = cv2.bitwise_not(img)
bit_not_rgb = cv2.cvtColor(bit_not, cv2.COLOR_BGR2RGB)

plt.imshow(bit_not_rgb)
plt.title("Bitwise NOT (Negative Image)")
plt.axis("off")
```

```
Out[32]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Bitwise NOT (Negative Image)



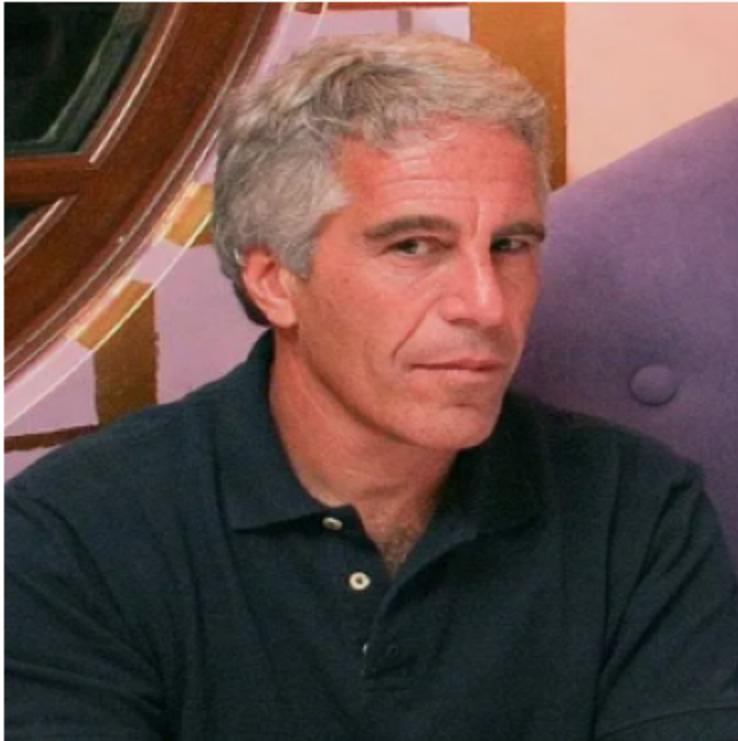
Image Resizing

```
In [33]: resized = cv2.resize(img, (300, 300), interpolation=cv2.INTER_LINEAR)
resized_rgb = cv2.cvtColor(resized, cv2.COLOR_BGR2RGB)

plt.imshow(resized_rgb)
plt.title("Resized Image")
plt.axis("off")
```

```
Out[33]: (np.float64(-0.5), np.float64(299.5), np.float64(299.5), np.float64(-0.5))
```

Resized Image



Rotation and Translation

```
In [34]: (h, w) = img.shape[:2]
center = (w//2, h//2)

matrix = cv2.getRotationMatrix2D(center, 45, 1.0)
rotated = cv2.warpAffine(img, matrix, (w, h))
rotated_rgb = cv2.cvtColor(rotated, cv2.COLOR_BGR2RGB)

plt.imshow(rotated_rgb)
plt.title("Rotated Image (45 Degrees)")
plt.axis("off")
```

Out[34]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))

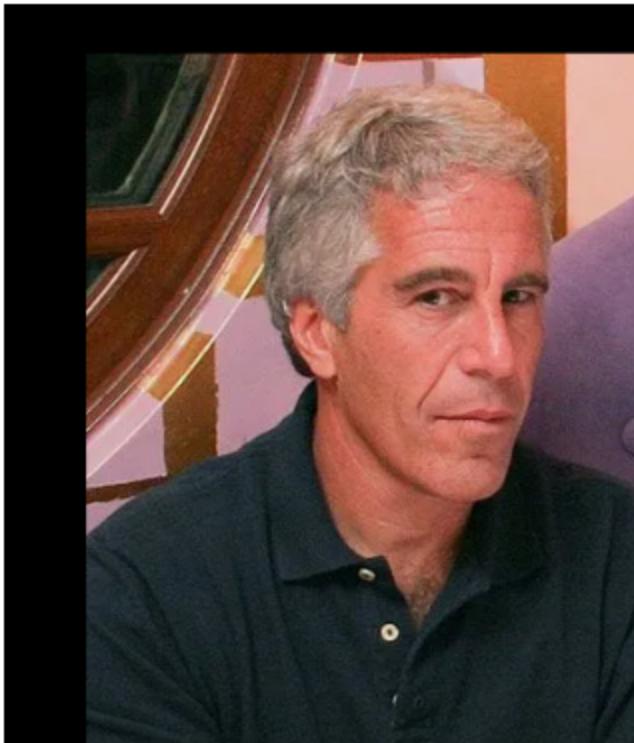
Rotated Image (45 Degrees)



```
In [35]: matrix_t = np.float32([[1, 0, 50],  
                           [0, 1, 30]])  
  
translated = cv2.warpAffine(img, matrix_t, (w, h))  
translated_rgb = cv2.cvtColor(translated, cv2.COLOR_BGR2RGB)  
  
plt.imshow(translated_rgb)  
plt.title("Translated Image")  
plt.axis("off")
```

Out[35]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))

Translated Image

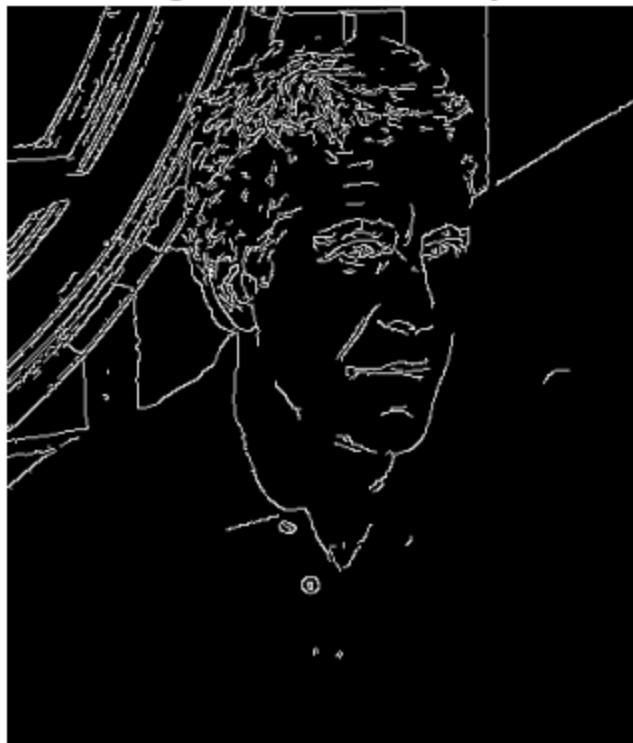


Additional Basic Operations

```
In [36]: edges = cv2.Canny(gray, 100, 200)
plt.imshow(edges, cmap='gray')
plt.title("Edge Detection (Canny)")
plt.axis("off")
```

```
Out[36]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Edge Detection (Canny)



```
In [37]: blur = cv2.GaussianBlur(img, (7,7), 0)
blur_rgb = cv2.cvtColor(blur, cv2.COLOR_BGR2RGB)

plt.imshow(blur_rgb)
plt.title("Gaussian Blurred Image")
plt.axis("off")
```

```
Out[37]: (np.float64(-0.5), np.float64(383.5), np.float64(449.5), np.float64(-0.5))
```

Gaussian Blurred Image



```
In [39]: img1 = np.zeros((256, 256), dtype=np.uint8)

for i in range(256):
    for j in range(256):
        if j > i:
            img1[i, j] = 255    # White
        else:
            img1[i, j] = 0      # Black

plt.imshow(img1, cmap='gray')
plt.title("Diagonal Half Black–White Image")
plt.axis("off")
```

```
Out[39]: (np.float64(-0.5), np.float64(255.5), np.float64(255.5), np.float64(-0.5))
```

Diagonal Half Black-White Image

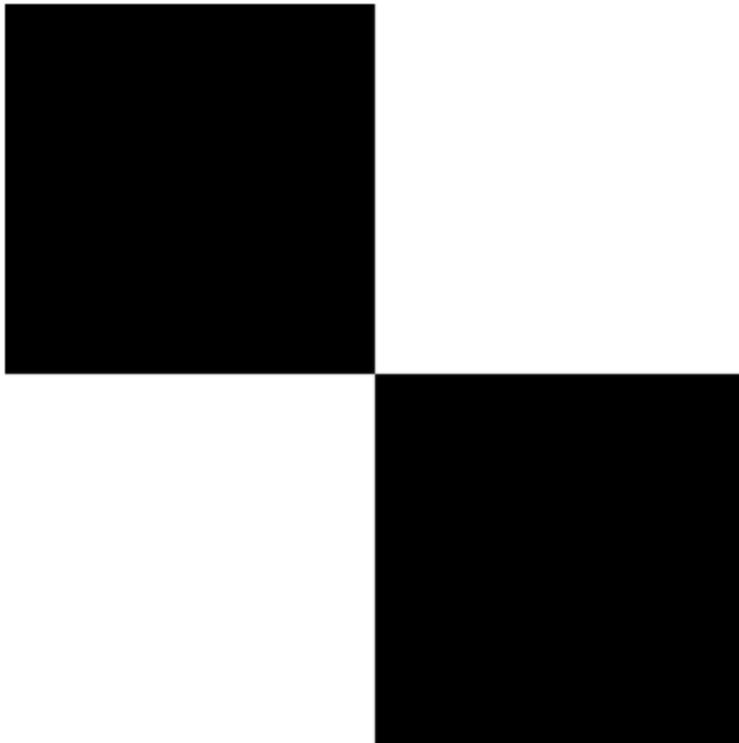


```
In [40]: img2 = np.zeros((256, 256), dtype=np.uint8)

img2[0:128, 0:128] = 0
img2[0:128, 128:256] = 255
img2[128:256, 0:128] = 255
img2[128:256, 128:256] = 0
plt.imshow(img2, cmap='gray')
plt.title("2x2 Quadrant Image")
plt.axis("off")
```

```
Out[40]: (np.float64(-0.5), np.float64(255.5), np.float64(255.5), np.float64(-0.5))
```

2x2 Quadrant Image



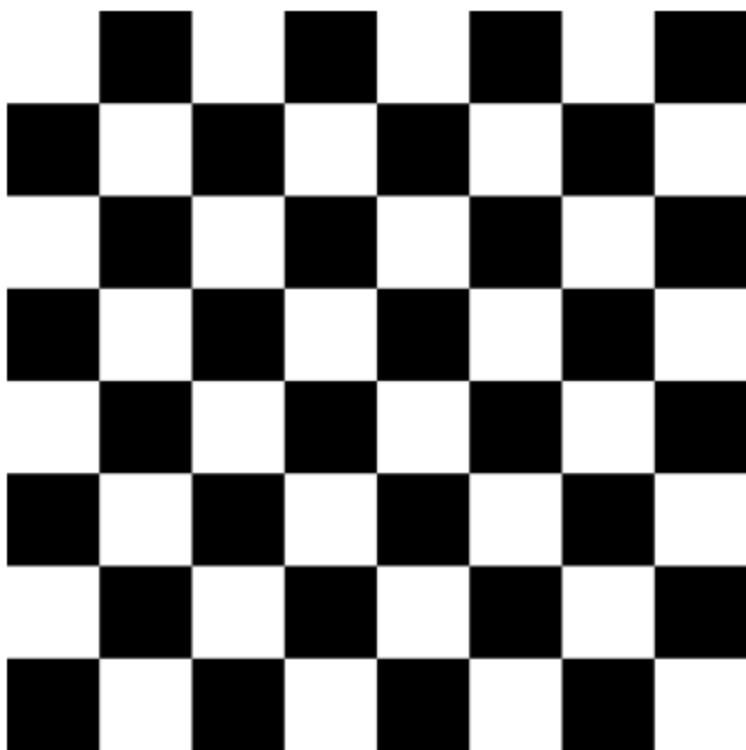
```
In [41]: img3 = np.zeros((256, 256), dtype=np.uint8)
block = 32

for i in range(0, 256, block):
    for j in range(0, 256, block):
        if (i//block + j//block) % 2 == 0:
            img3[i:i+block, j:j+block] = 255

plt.imshow(img3, cmap='gray')
plt.title("Chessboard Pattern")
plt.axis("off")
```

```
Out[41]: (np.float64(-0.5), np.float64(255.5), np.float64(255.5), np.float64(-0.5))
```

Chessboard Pattern



Algorithm: Piecewise Linear Transformation

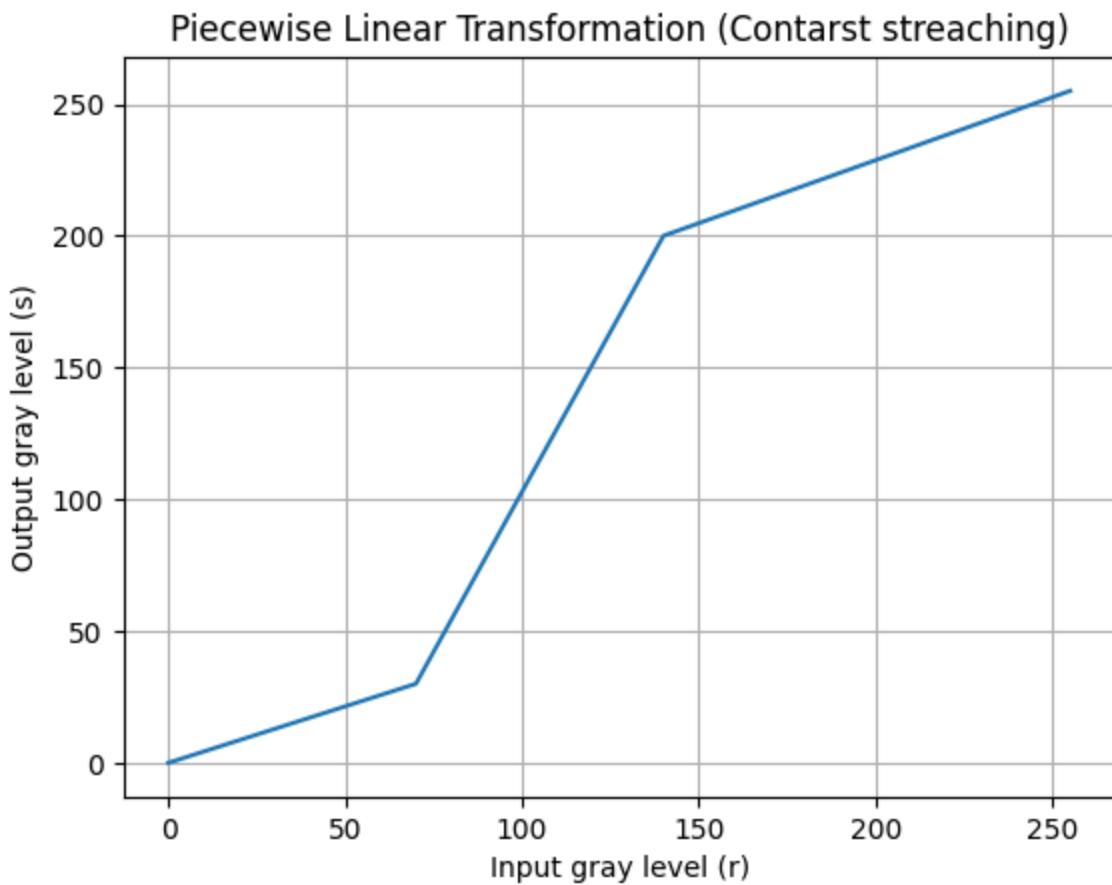
```
In [67]: import numpy as np  
import matplotlib.pyplot as plt
```

```
In [68]: r = np.arange(256)  
s = np.zeros(256)
```

```
In [69]: r1 , r2 = 70, 140  
s1 , s2 = 30 , 200
```

```
In [70]: for i in range(256):  
    if r[i] < r1:  
        s[i] = (s1/r1) * r[i]  
    elif r[i] <= r2:  
        s[i] = ((s2 - s1)/(r2 - r1)) * (r[i] - r1) + s1  
    else:  
        s[i] = ((255 - s2)/(255 - r2)) * (r[i] - r2) + s2
```

```
In [71]: plt.plot(r, s)  
plt.xlabel("Input gray level (r)")  
plt.ylabel("Output gray level (s)")  
plt.title("Piecewise Linear Transformation (Contrast stretching)")  
plt.grid()  
plt.show()
```



```
In [72]: import cv2
import numpy as np
import matplotlib.pyplot as plt
```

```
In [ ]: img = cv2.imread(
    r"D:\MIT WPU B.Tech Data\3rd Year Data\IPPR\puppy-with-gray-scale-noise.we
    b
)
```

Piecewise Linear Transformation (ON IMAGE)

```
In [83]: r_vals = np.arange(256)
s_vals = np.zeros(256)

r1, r2 = 70, 140
s1, s2 = 30, 200

for i in range(256):
    if r_vals[i] < r1:
        s_vals[i] = (s1 / r1) * r_vals[i]
    elif r_vals[i] <= r2:
```

```

        s_vals[i] = ((s2 - s1) / (r2 - r1)) * (r_vals[i] - r1) + s1
    else:
        s_vals[i] = ((255 - s2) / (255 - r2)) * (r_vals[i] - r2) + s2

plt.figure(figsize=(15, 4))

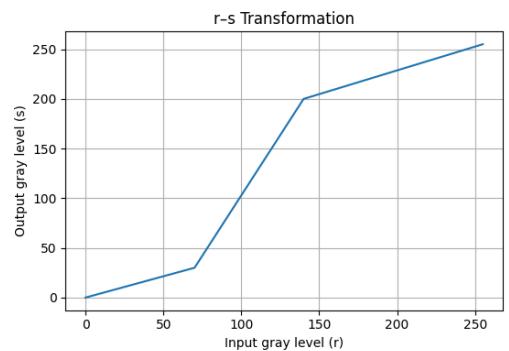
plt.subplot(1, 3, 1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(1, 3, 2)
plt.imshow(piecewise, cmap='gray')
plt.title("Piecewise Linear Output")
plt.axis("off")

plt.subplot(1, 3, 3)
plt.plot(r_vals, s_vals)
plt.xlabel("Input gray level (r)")
plt.ylabel("Output gray level (s)")
plt.title("r-s Transformation")
plt.grid()

plt.tight_layout()
plt.show()

```



Gray Level Slicing WITHOUT Background

```

In [85]: r_vals = np.arange(256)
s_vals = np.zeros(256)

r1, r2 = 100, 150

for i in range(256):

```

```

if r1 <= r_vals[i] <= r2:
    s_vals[i] = 255
else:
    s_vals[i] = 0
plt.figure(figsize=(15, 4))

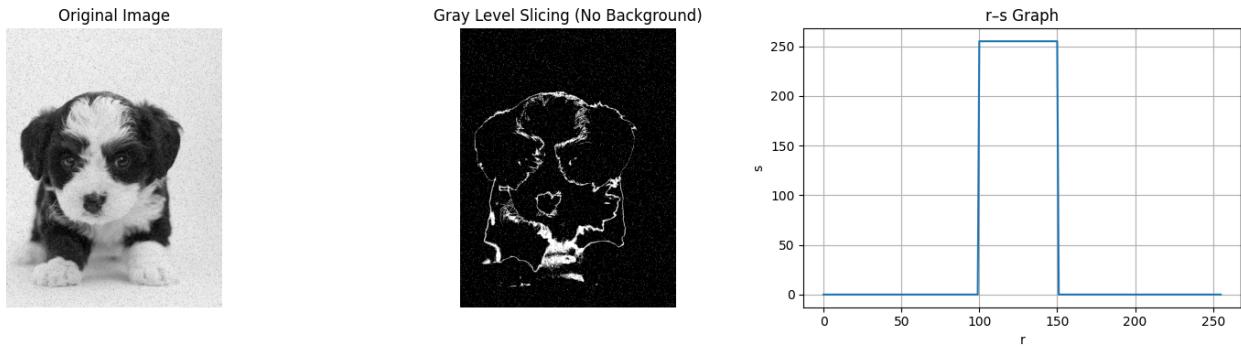
plt.subplot(1, 3, 1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(1, 3, 2)
plt.imshow(slice_no_bg, cmap='gray')
plt.title("Gray Level Slicing (No Background)")
plt.axis("off")

plt.subplot(1, 3, 3)
plt.plot(r_vals, s_vals)
plt.xlabel("r")
plt.ylabel("s")
plt.title("r-s Graph")
plt.grid()

plt.tight_layout()
plt.show()

```



Gray Level Slicing WITH Background

```

In [86]: r_vals = np.arange(256)
s_vals = np.zeros(256)

for i in range(256):
    if r1 <= r_vals[i] <= r2:
        s_vals[i] = 255
    else:
        s_vals[i] = r_vals[i]
plt.figure(figsize=(15, 4))

plt.subplot(1, 3, 1)
plt.imshow(img, cmap='gray')

```

```

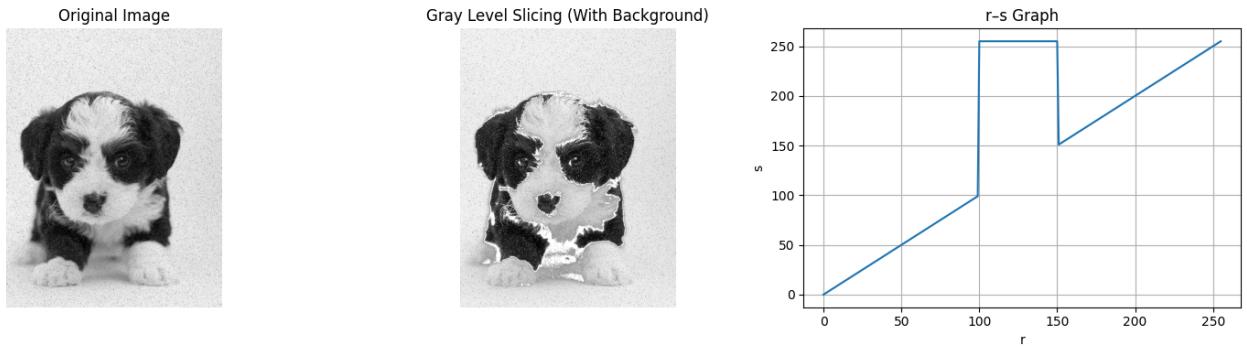
plt.title("Original Image")
plt.axis("off")

plt.subplot(1, 3, 2)
plt.imshow(slice_with_bg, cmap='gray')
plt.title("Gray Level Slicing (With Background)")
plt.axis("off")

plt.subplot(1, 3, 3)
plt.plot(r_vals, s_vals)
plt.xlabel("r")
plt.ylabel("s")
plt.title("r-s Graph")
plt.grid()

plt.tight_layout()
plt.show()

```



Contrast Stretching

```

In [87]: r_vals = np.arange(256)
s_vals = ((r_vals - r_min) / (r_max - r_min)) * 255
plt.figure(figsize=(15, 4))

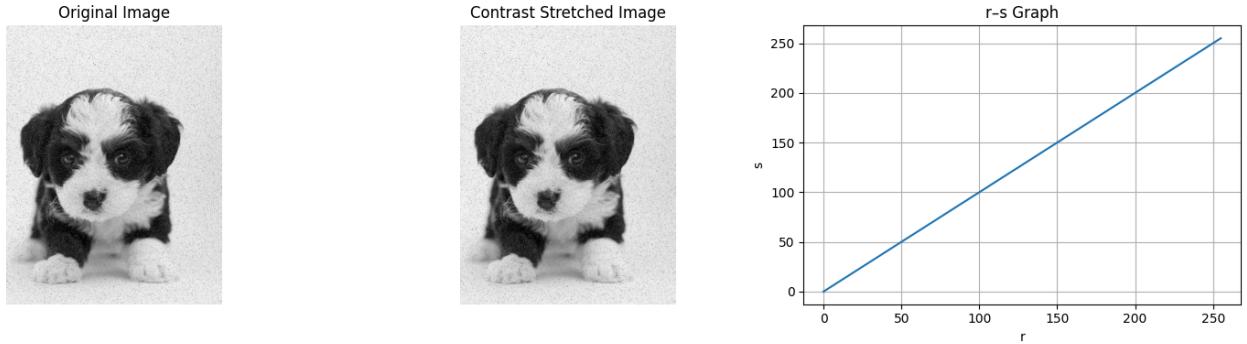
plt.subplot(1, 3, 1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(1, 3, 2)
plt.imshow(contrast_stretch, cmap='gray')
plt.title("Contrast Stretched Image")
plt.axis("off")

plt.subplot(1, 3, 3)
plt.plot(r_vals, s_vals)
plt.xlabel("r")
plt.ylabel("s")
plt.title("r-s Graph")
plt.grid()

```

```
plt.tight_layout()  
plt.show()
```



Thresholding

```
In [88]: T = 120  
r_vals = np.arange(256)  
s_vals = np.where(r_vals >= T, 255, 0)  
plt.figure(figsize=(15, 4))  
  
plt.subplot(1, 3, 1)  
plt.imshow(img, cmap='gray')  
plt.title("Original Image")  
plt.axis("off")  
  
plt.subplot(1, 3, 2)  
plt.imshow(threshold_img, cmap='gray')  
plt.title("Thresholded Image")  
plt.axis("off")  
  
plt.subplot(1, 3, 3)  
plt.plot(r_vals, s_vals)  
plt.xlabel("r")  
plt.ylabel("s")  
plt.title("r-s Graph")  
plt.grid()  
  
plt.tight_layout()  
plt.show()
```

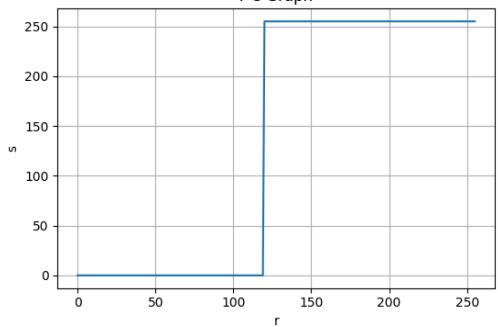
Original Image

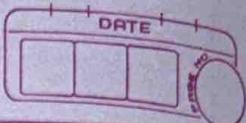


Thresholded Image



r-s Graph





IPPR

* Exp - 1

* Post lab Questions.

- 1) What do you mean by gray level?
→ In image processing, a grey level refers to the intensity value of a pixel in a grayscale image. It tells you how bright or dark pixel is.

- 2) Write the expression to find the number of bits to store a digital image.
→ The expression to store digital image using number of bits is.

$$\text{Num of bits} = M \times N \times K$$

M - number of rows.

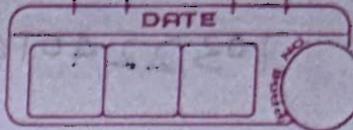
N - number of columns.

K - number of bits per pixel (bpp)

- 3) Name types of resolutions w.r.t a digital image.

→ The main types of resolution w.r.t a digital image are.

- 1) Spatial resolution
- 2) Grey level (intensity) Resolution
- 3) Temporal resolution.
- 4) Physical level resolution.



4) Specify the elements of the DIP system.

→ A typical Digital Image processing system includes.

1) Image Acquisition

2) Image Storage

3) Image Processing

4) Image Display

5) Image Transmission

6) Image Interpretation

5) Write any four applications of DIP

→ Common application of DIP are.

1) Medical Imaging (e.g., CT, MRI)

2) Remote sensing (satellite)

3) Robotics & Machine vision

4) Bio metrics (Face, fingerprint).

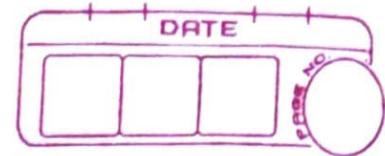


Image
Display

Computer

Mass
storage

Hard
copy
Device

Image processing
hardware

Image
processing
software

Image
Sensor

Problem - Domain.

Part 1

```
In [43]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import random
```

```
In [44]: import cv2
import matplotlib.pyplot as plt

img = cv2.imread(
    r"D:\MIT WPU B.Tech Data\3rd Year Data\IPPR\Image-Processing-and-Pattern-R
    cv2.IMREAD_GRAYSCALE
)

if img is None:
    print("Image not found or path is incorrect!")
else:
    print("Image loaded successfully!")
    plt.imshow(img, cmap='gray')
    plt.title("Original Grayscale Image")
    plt.axis("off")
```

Image loaded successfully!

Original Grayscale Image



Gaussian noise

```
In [45]: mean = 0
std = 25
```

```
gaussian_noise = np.random.normal(mean, std, img.shape)
noisy_img = img.astype(np.float32) + gaussian_noise
noisy_img = np.clip(noisy_img, 0, 255).astype(np.uint8)

# Display noisy image
plt.imshow(noisy_img, cmap='gray')
plt.title("Image with Gaussian Noise")
plt.axis("off")
```

Out[45]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Image with Gaussian Noise



Add Salt & Pepper Noise

```
In [46]: salt_pepper_noisy = img.copy()
prob = 0.02

rand = np.random.rand(*img.shape)

salt_pepper_noisy[rand < prob/2] = 0
salt_pepper_noisy[rand > 1 - prob/2] = 255

plt.imshow(salt_pepper_noisy, cmap='gray')
plt.title("Salt & Pepper Noise")
plt.axis("off")
```

Out[46]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Salt & Pepper Noise



Add Uniform Noise

```
In [47]: low = -30
high = 30

uniform_noise = np.random.uniform(low, high, img.shape)
uniform_noisy = img + uniform_noise

uniform_noisy = np.clip(uniform_noisy, 0, 255).astype(np.uint8)

plt.imshow(uniform_noisy, cmap='gray')
plt.title("Uniform Noise")
plt.axis("off")
```

Out[47]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Uniform Noise



```
In [48]: import cv2  
  
kernel_size = 5  
avg_filtered_img = cv2.blur(noisy_img, (kernel_size, kernel_size))  
import matplotlib.pyplot as plt  
  
plt.imshow(avg_filtered_img, cmap='gray')  
plt.title("Averaging Filtered Image")  
plt.axis("off")
```

```
Out[48]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))
```

Averaging Filtered Image



Apply avg filter

```
In [49]: kernel_size = 5

gaussian_avg = cv2.blur(gaussian_noise, (kernel_size, kernel_size))
salt_pepper_avg = cv2.blur(salt_pepper_noisy, (kernel_size, kernel_size))
uniform_avg = cv2.blur(uniform_noisy, (kernel_size, kernel_size))
```

```
In [50]: plt.figure(figsize=(15,8))

plt.subplot(3,3,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(3,3,2)
plt.imshow(gaussian_noise, cmap='gray')
plt.title("Gaussian Noise")
plt.axis("off")

plt.subplot(3,3,3)
plt.imshow(gaussian_avg, cmap='gray')
plt.title("Gaussian + Averaging")
plt.axis("off")

plt.subplot(3,3,4)
plt.imshow(salt_pepper_noisy, cmap='gray')
plt.title("Salt & Pepper Noise")
plt.axis("off")
```

```

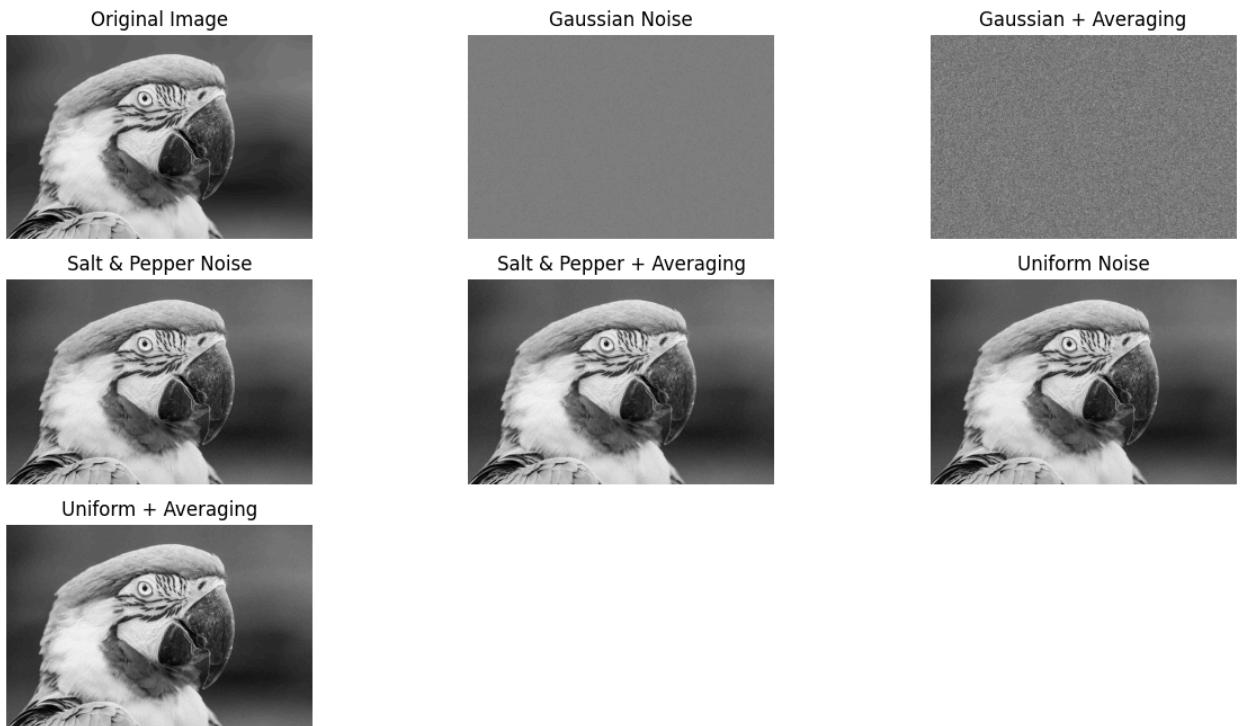
plt.subplot(3,3,5)
plt.imshow(salt_pepper_avg, cmap='gray')
plt.title("Salt & Pepper + Averaging")
plt.axis("off")

plt.subplot(3,3,6)
plt.imshow(uniform_noisy, cmap='gray')
plt.title("Uniform Noise")
plt.axis("off")

plt.subplot(3,3,7)
plt.imshow(uniform_avg, cmap='gray')
plt.title("Uniform + Averaging")
plt.axis("off")

plt.show()

```



Compare Median Filtered Image with Averaging Filtered Image

```

In [51]: median_filtered_img = cv2.medianBlur(noisy_img, 5)

plt.figure(figsize=(12,4))

plt.subplot(1,3,1)
plt.imshow(noisy_img, cmap='gray')
plt.title("Noisy Image")
plt.axis("off")

```

```

plt.subplot(1,3,2)
plt.imshow(avg_filtered_img, cmap='gray')
plt.title("Averaging Filtered Image")
plt.axis("off")

plt.subplot(1,3,3)
plt.imshow(median_filtered_img, cmap='gray')
plt.title("Median Filtered Image")
plt.axis("off")

plt.show()

```



Masking

```

In [52]: mask_sizes = [5, 7, 9, 11, 13, 15]

plt.figure(figsize=(15,8))

plt.subplot(2,4,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

for i, k in enumerate(mask_sizes):
    smoothed = cv2.blur(img, (k, k))

    plt.subplot(2,4,i+2)
    plt.imshow(smoothed, cmap='gray')
    plt.title(f"Averaging {k}x{k}")
    plt.axis("off")

plt.show()

```



Apply Nonlinear Filters (Median, Min, Max)

```
In [53]: import cv2
import matplotlib.pyplot as plt
import numpy as np

k = 5

median_img = cv2.medianBlur(img, k)

kernel = np.ones((k, k), np.uint8)
min_img = cv2.erode(img, kernel)

max_img = cv2.dilate(img, kernel)
plt.figure(figsize=(12,6))

plt.subplot(2,2,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(2,2,2)
plt.imshow(median_img, cmap='gray')
plt.title("Median Filtered Image")
plt.axis("off")

plt.subplot(2,2,3)
plt.imshow(min_img, cmap='gray')
plt.title("Min Filtered Image")
plt.axis("off")
```

```

plt.subplot(2,2,4)
plt.imshow(max_img, cmap='gray')
plt.title("Max Filtered Image")
plt.axis("off")

plt.show()

```

Original Image



Median Filtered Image



Min Filtered Image



Max Filtered Image



In [54]:

```

import cv2
import numpy as np
import matplotlib.pyplot as plt

```

```

mean = 0
std = 25
gaussian_noise = np.random.normal(mean, std, img.shape)
gaussian_noisy = img.astype(np.float32) + gaussian_noise
gaussian_noisy = np.clip(gaussian_noisy, 0, 255).astype(np.uint8)

salt_pepper_noisy = img.copy()
prob = 0.02
rand = np.random.rand(*img.shape)
salt_pepper_noisy[rand < prob/2] = 0
salt_pepper_noisy[rand > 1 - prob/2] = 255

uniform_noise = np.random.uniform(-30, 30, img.shape)
uniform_noisy = img.astype(np.float32) + uniform_noise
uniform_noisy = np.clip(uniform_noisy, 0, 255).astype(np.uint8)

```

```
In [56]: k = 5
kernel = np.ones((k, k), np.uint8)

gaussian_median = cv2.medianBlur(gaussian_noisy, k)
salt_pepper_median = cv2.medianBlur(salt_pepper_noisy, k)
uniform_median = cv2.medianBlur(uniform_noisy, k)

gaussian_min = cv2.erode(gaussian_noisy, kernel)
salt_pepper_min = cv2.erode(salt_pepper_noisy, kernel)
uniform_min = cv2.erode(uniform_noisy, kernel)

gaussian_max = cv2.dilate(gaussian_noisy, kernel)
salt_pepper_max = cv2.dilate(salt_pepper_noisy, kernel)
uniform_max = cv2.dilate(uniform_noisy, kernel)

plt.figure(figsize=(12,14))

plt.subplot(5,3,1); plt.imshow(img, cmap='gray'); plt.title("Original"); plt.a
plt.subplot(5,3,2); plt.imshow(img, cmap='gray'); plt.title("Original"); plt.a
plt.subplot(5,3,3); plt.imshow(img, cmap='gray'); plt.title("Original"); plt.a

plt.subplot(5,3,4); plt.imshow(gaussian_noisy, cmap='gray'); plt.title("Gaussi
plt.subplot(5,3,5); plt.imshow(salt_pepper_noisy, cmap='gray'); plt.title("Sal
plt.subplot(5,3,6); plt.imshow(uniform_noisy, cmap='gray'); plt.title("Uniform

plt.subplot(5,3,7); plt.imshow(gaussian_median, cmap='gray'); plt.title("Gaussi
plt.subplot(5,3,8); plt.imshow(salt_pepper_median, cmap='gray'); plt.title("S&P
plt.subplot(5,3,9); plt.imshow(uniform_median, cmap='gray'); plt.title("Uniform

plt.subplot(5,3,10); plt.imshow(gaussian_min, cmap='gray'); plt.title("Gaussia
plt.subplot(5,3,11); plt.imshow(salt_pepper_min, cmap='gray'); plt.title("S&P
plt.subplot(5,3,12); plt.imshow(uniform_min, cmap='gray'); plt.title("Uniform

plt.subplot(5,3,13); plt.imshow(gaussian_max, cmap='gray'); plt.title("Gaussia
plt.subplot(5,3,14); plt.imshow(salt_pepper_max, cmap='gray'); plt.title("S&P
plt.subplot(5,3,15); plt.imshow(uniform_max, cmap='gray'); plt.title("Uniform

plt.show()
```



Part 2

```
In [ ]: import cv2
import matplotlib.pyplot as plt
img = cv2.imread(
    r"D:\MIT WPU B.Tech Data\3nd Year Data\IPPR\Image-Processing-and-Pattern-R
    cv2.IMREAD_GRAYSCALE
```

```
)  
  
if img is None:  
    print("Image not found!")  
else:  
    plt.imshow(img, cmap='gray')  
    plt.title("Original Grayscale Image")  
    plt.axis("off")
```

Original Grayscale Image



Apply Sharpening Filter on the Given Image

In [58]:

```
import numpy as np  
  
sharpen_kernel = np.array([[ 0, -1,  0],  
                          [-1,  5, -1],  
                          [ 0, -1,  0]])  
  
sharpened_img = cv2.filter2D(img, -1, sharpen_kernel)  
  
plt.imshow(sharpened_img, cmap='gray')  
plt.title("Sharpened Image")  
plt.axis("off")
```

Out[58]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Sharpened Image



```
In [59]: plt.figure(figsize=(10,4))

plt.subplot(1,2,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(1,2,2)
plt.imshow(sharpened_img, cmap='gray')
plt.title("Sharpened Image")
plt.axis("off")

plt.show()
```

Original Image



Sharpened Image



Apply sharpening filter on the noisy image.

```
In [61]: import numpy as np
import cv2
import matplotlib.pyplot as plt

sharpen_kernel = np.array([[ 0, -1,  0],
                          [-1,  5, -1],
                          [ 0, -1,  0]])
sharpened_noisy_img = cv2.filter2D(noisy_img, -1, sharpen_kernel)
plt.figure(figsize=(10,4))

plt.subplot(1,2,1)
plt.imshow(noisy_img, cmap='gray')
plt.title("Noisy Image")
plt.axis("off")

plt.subplot(1,2,2)
plt.imshow(sharpened_noisy_img, cmap='gray')
plt.title("Sharpened Noisy Image")
plt.axis("off")

plt.show()
```

Noisy Image



Sharpened Noisy Image



Display original and sharpened images.

```
In [62]: import matplotlib.pyplot as plt

plt.figure(figsize=(8,4))

plt.subplot(1,2,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")
```

```
plt.subplot(1,2,2)
plt.imshow(sharpened_img, cmap='gray')
plt.title("Sharpened Image")
plt.axis("off")

plt.show()
```

Original Image



Sharpened Image



Mean Squared Error (MSE)

```
In [63]: import numpy as np

mse = np.mean((img.astype(np.float32) - sharpened_img.astype(np.float32)) ** 2)

print("Mean Squared Error (MSE):", mse)
```

Mean Squared Error (MSE): 340.6078

Peak Signal-to-Noise Ratio (PSNR)

```
In [64]: import math

if mse == 0:
    psnr = float('inf')
else:
    max_pixel = 255.0
    psnr = 20 * math.log10(max_pixel / math.sqrt(mse))

print("Peak Signal-to-Noise Ratio (PSNR):", psnr, "dB")
```

Peak Signal-to-Noise Ratio (PSNR): 22.808257868711635 dB



IPPR

Name: Uzair Shaikh
PRN : 1032240100.

Jh
6/2

* Experiment -3

* Post lab Questions.

1) Write a note on sharpening Filters.

→ Sharpening filters are image processing techniques used to enhance edge and fine details in an image, making it appear clearer and more defined. They work by emphasizing high-frequency components, which represent rapid changes in intensity such as edges. Common sharpening methods includes laplacian, sobel, Prewitt, and unsharp masking, each used to highlight important image feature. Sharpening filters are widely applied in medical imaging, remote sensing, and digital photography, but excessive use can amplify noise and cause visual artifacts.

Advantages:

- 1) Enhances edges and find details
- 2) Improves visual quality.
- 3) Useful for feature extraction.

Disadvantages:

- 1) Amplifies Noise
- 2) over-sharpening can cause artifacts and unnatural apperence.

5.

Example:-

10	10	210	210	210	MASK
10	10	210	210	210	0 -1 0
10	10	210	210	210	-1 4 -1
10	10	210	210	210	0 1 -1 0
10	10	210	210	210	

Pixel (row 2, col 2)

$$= (4 \times 10) + (10 \times (-1)) + (10 \times (-1)) + (10 \times (-1)) + (210 \times (-1))$$

$$= -200$$

new image

10	10	210	210	210
10	-200	200	0	210
10	-200	200	0	210
10	-200	200	0	210
10	10	210	210	210

2. Consider an input image row $[4 \ 3 \ 2 \ 1]$ with intensity values in range 0 to 15. Determine negative of the image.

$$\rightarrow [4 \ 3 \ 2 \ 1]$$

$$\text{Negative pixel} = (L-1) - f(n, y)$$

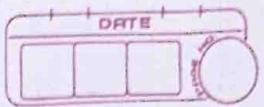
$$\text{For } 4 \rightarrow 15-4 = 11$$

$$\text{For } 3 \rightarrow 15-3 = 12$$

$$\text{For } 2 \rightarrow 15-2 = 13$$

$$\text{For } 1 \rightarrow 15-1 = 14$$

new row



[11 12 13 14]

3. Determine the following value of the central pixel of the following image by applying a 3×3 size

\rightarrow

10	11	11
10	255	11
12	12	11

a) Mean filter

$$\text{Mean} = \frac{1}{n} \sum f(x_i, y_i)$$

$$= \frac{10+10+12+11+11+11+12+255}{9}$$

$$= 38.11 \approx 38$$

new image

10	11	11
10	38	11
12	12	11

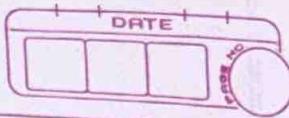
b) Median filter

10, 10, 11, 11, 11, 11, 12, 12, 255

median = 11

new image:

10	11	11
10	11	11
12	12	11



c) Mode filter

count →

- 10 → 2 times
- 11 → 4 times
- 12 → 2 times
- 255 → 1 times

most frequent → 11

10	11	11
10	11	11
12	12	11



Part 1

```
In [43]: import cv2
import numpy as np
import matplotlib.pyplot as plt
import random
```

```
In [44]: import cv2
import matplotlib.pyplot as plt

img = cv2.imread(
    r"D:\MIT WPU B.Tech Data\3rd Year Data\IPPR\Image-Processing-and-Pattern-R
    cv2.IMREAD_GRAYSCALE
)

if img is None:
    print("Image not found or path is incorrect!")
else:
    print("Image loaded successfully!")
    plt.imshow(img, cmap='gray')
    plt.title("Original Grayscale Image")
    plt.axis("off")
```

Image loaded successfully!

Original Grayscale Image



Gaussian noise

```
In [45]: mean = 0
std = 25
```

```
gaussian_noise = np.random.normal(mean, std, img.shape)
noisy_img = img.astype(np.float32) + gaussian_noise
noisy_img = np.clip(noisy_img, 0, 255).astype(np.uint8)

# Display noisy image
plt.imshow(noisy_img, cmap='gray')
plt.title("Image with Gaussian Noise")
plt.axis("off")
```

Out[45]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Image with Gaussian Noise



Add Salt & Pepper Noise

```
In [46]: salt_pepper_noisy = img.copy()
prob = 0.02

rand = np.random.rand(*img.shape)

salt_pepper_noisy[rand < prob/2] = 0
salt_pepper_noisy[rand > 1 - prob/2] = 255

plt.imshow(salt_pepper_noisy, cmap='gray')
plt.title("Salt & Pepper Noise")
plt.axis("off")
```

Out[46]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Salt & Pepper Noise



Add Uniform Noise

```
In [47]: low = -30
high = 30

uniform_noise = np.random.uniform(low, high, img.shape)
uniform_noisy = img + uniform_noise

uniform_noisy = np.clip(uniform_noisy, 0, 255).astype(np.uint8)

plt.imshow(uniform_noisy, cmap='gray')
plt.title("Uniform Noise")
plt.axis("off")
```

Out[47]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Uniform Noise



```
In [48]: import cv2  
  
kernel_size = 5  
avg_filtered_img = cv2.blur(noisy_img, (kernel_size, kernel_size))  
import matplotlib.pyplot as plt  
  
plt.imshow(avg_filtered_img, cmap='gray')  
plt.title("Averaging Filtered Image")  
plt.axis("off")
```

```
Out[48]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))
```

Averaging Filtered Image



Apply avg filter

```
In [49]: kernel_size = 5

gaussian_avg = cv2.blur(gaussian_noise, (kernel_size, kernel_size))
salt_pepper_avg = cv2.blur(salt_pepper_noisy, (kernel_size, kernel_size))
uniform_avg = cv2.blur(uniform_noisy, (kernel_size, kernel_size))
```

```
In [50]: plt.figure(figsize=(15,8))

plt.subplot(3,3,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(3,3,2)
plt.imshow(gaussian_noise, cmap='gray')
plt.title("Gaussian Noise")
plt.axis("off")

plt.subplot(3,3,3)
plt.imshow(gaussian_avg, cmap='gray')
plt.title("Gaussian + Averaging")
plt.axis("off")

plt.subplot(3,3,4)
plt.imshow(salt_pepper_noisy, cmap='gray')
plt.title("Salt & Pepper Noise")
plt.axis("off")
```

```

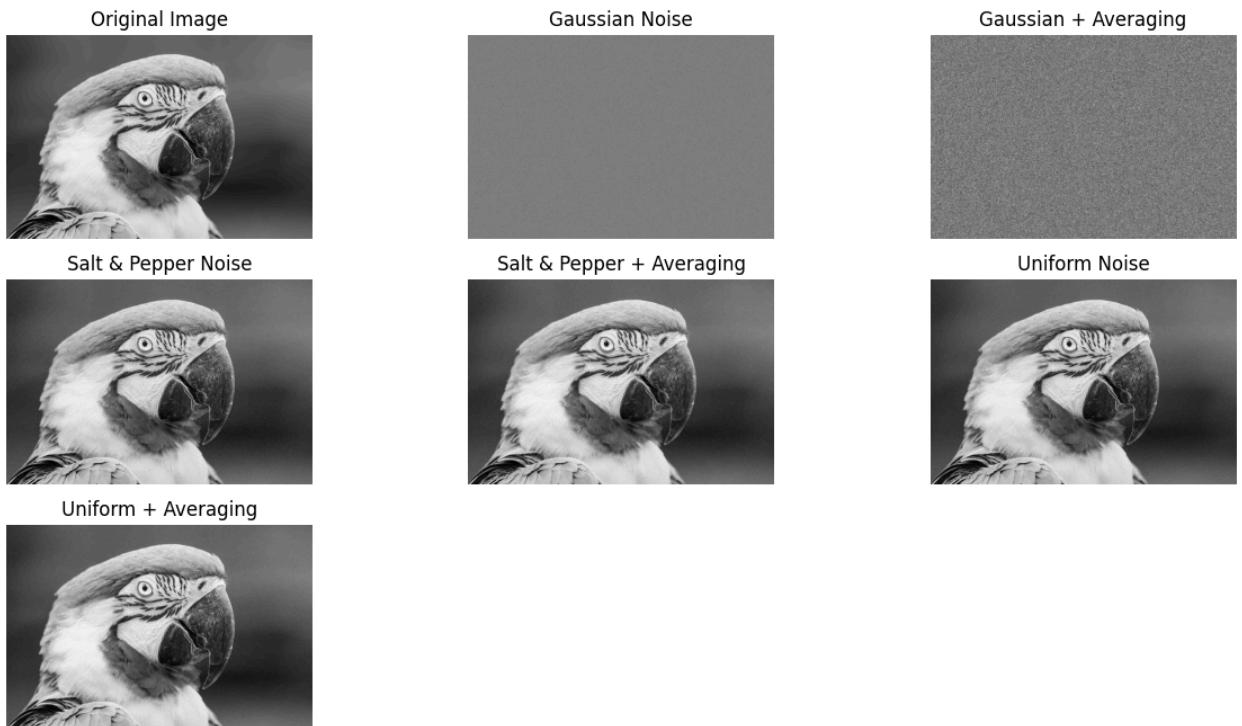
plt.subplot(3,3,5)
plt.imshow(salt_pepper_avg, cmap='gray')
plt.title("Salt & Pepper + Averaging")
plt.axis("off")

plt.subplot(3,3,6)
plt.imshow(uniform_noisy, cmap='gray')
plt.title("Uniform Noise")
plt.axis("off")

plt.subplot(3,3,7)
plt.imshow(uniform_avg, cmap='gray')
plt.title("Uniform + Averaging")
plt.axis("off")

plt.show()

```



Compare Median Filtered Image with Averaging Filtered Image

```

In [51]: median_filtered_img = cv2.medianBlur(noisy_img, 5)

plt.figure(figsize=(12,4))

plt.subplot(1,3,1)
plt.imshow(noisy_img, cmap='gray')
plt.title("Noisy Image")
plt.axis("off")

```

```

plt.subplot(1,3,2)
plt.imshow(avg_filtered_img, cmap='gray')
plt.title("Averaging Filtered Image")
plt.axis("off")

plt.subplot(1,3,3)
plt.imshow(median_filtered_img, cmap='gray')
plt.title("Median Filtered Image")
plt.axis("off")

plt.show()

```



Masking

```

In [52]: mask_sizes = [5, 7, 9, 11, 13, 15]

plt.figure(figsize=(15,8))

plt.subplot(2,4,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

for i, k in enumerate(mask_sizes):
    smoothed = cv2.blur(img, (k, k))

    plt.subplot(2,4,i+2)
    plt.imshow(smoothed, cmap='gray')
    plt.title(f"Averaging {k}x{k}")
    plt.axis("off")

plt.show()

```



Apply Nonlinear Filters (Median, Min, Max)

```
In [53]: import cv2
import matplotlib.pyplot as plt
import numpy as np

k = 5

median_img = cv2.medianBlur(img, k)

kernel = np.ones((k, k), np.uint8)
min_img = cv2.erode(img, kernel)

max_img = cv2.dilate(img, kernel)
plt.figure(figsize=(12,6))

plt.subplot(2,2,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(2,2,2)
plt.imshow(median_img, cmap='gray')
plt.title("Median Filtered Image")
plt.axis("off")

plt.subplot(2,2,3)
plt.imshow(min_img, cmap='gray')
plt.title("Min Filtered Image")
plt.axis("off")
```

```

plt.subplot(2,2,4)
plt.imshow(max_img, cmap='gray')
plt.title("Max Filtered Image")
plt.axis("off")

plt.show()

```

Original Image



Median Filtered Image



Min Filtered Image



Max Filtered Image



In [54]:

```

import cv2
import numpy as np
import matplotlib.pyplot as plt

```

```

mean = 0
std = 25
gaussian_noise = np.random.normal(mean, std, img.shape)
gaussian_noisy = img.astype(np.float32) + gaussian_noise
gaussian_noisy = np.clip(gaussian_noisy, 0, 255).astype(np.uint8)

salt_pepper_noisy = img.copy()
prob = 0.02
rand = np.random.rand(*img.shape)
salt_pepper_noisy[rand < prob/2] = 0
salt_pepper_noisy[rand > 1 - prob/2] = 255

uniform_noise = np.random.uniform(-30, 30, img.shape)
uniform_noisy = img.astype(np.float32) + uniform_noise
uniform_noisy = np.clip(uniform_noisy, 0, 255).astype(np.uint8)

```

```
In [56]: k = 5
kernel = np.ones((k, k), np.uint8)

gaussian_median = cv2.medianBlur(gaussian_noisy, k)
salt_pepper_median = cv2.medianBlur(salt_pepper_noisy, k)
uniform_median = cv2.medianBlur(uniform_noisy, k)

gaussian_min = cv2.erode(gaussian_noisy, kernel)
salt_pepper_min = cv2.erode(salt_pepper_noisy, kernel)
uniform_min = cv2.erode(uniform_noisy, kernel)

gaussian_max = cv2.dilate(gaussian_noisy, kernel)
salt_pepper_max = cv2.dilate(salt_pepper_noisy, kernel)
uniform_max = cv2.dilate(uniform_noisy, kernel)

plt.figure(figsize=(12,14))

plt.subplot(5,3,1); plt.imshow(img, cmap='gray'); plt.title("Original"); plt.a
plt.subplot(5,3,2); plt.imshow(img, cmap='gray'); plt.title("Original"); plt.a
plt.subplot(5,3,3); plt.imshow(img, cmap='gray'); plt.title("Original"); plt.a

plt.subplot(5,3,4); plt.imshow(gaussian_noisy, cmap='gray'); plt.title("Gaussi
plt.subplot(5,3,5); plt.imshow(salt_pepper_noisy, cmap='gray'); plt.title("Sal
plt.subplot(5,3,6); plt.imshow(uniform_noisy, cmap='gray'); plt.title("Uniform

plt.subplot(5,3,7); plt.imshow(gaussian_median, cmap='gray'); plt.title("Gaussi
plt.subplot(5,3,8); plt.imshow(salt_pepper_median, cmap='gray'); plt.title("S&P
plt.subplot(5,3,9); plt.imshow(uniform_median, cmap='gray'); plt.title("Uniform

plt.subplot(5,3,10); plt.imshow(gaussian_min, cmap='gray'); plt.title("Gaussia
plt.subplot(5,3,11); plt.imshow(salt_pepper_min, cmap='gray'); plt.title("S&P
plt.subplot(5,3,12); plt.imshow(uniform_min, cmap='gray'); plt.title("Uniform

plt.subplot(5,3,13); plt.imshow(gaussian_max, cmap='gray'); plt.title("Gaussia
plt.subplot(5,3,14); plt.imshow(salt_pepper_max, cmap='gray'); plt.title("S&P
plt.subplot(5,3,15); plt.imshow(uniform_max, cmap='gray'); plt.title("Uniform

plt.show()
```



Part 2

```
In [ ]: import cv2
import matplotlib.pyplot as plt
img = cv2.imread(
    r"D:\MIT WPU B.Tech Data\3nd Year Data\IPPR\Image-Processing-and-Pattern-R
    cv2.IMREAD_GRAYSCALE
```

```
)  
  
if img is None:  
    print("Image not found!")  
else:  
    plt.imshow(img, cmap='gray')  
    plt.title("Original Grayscale Image")  
    plt.axis("off")
```

Original Grayscale Image



Apply Sharpening Filter on the Given Image

In [58]:

```
import numpy as np  
  
sharpen_kernel = np.array([[ 0, -1,  0],  
                          [-1,  5, -1],  
                          [ 0, -1,  0]])  
  
sharpened_img = cv2.filter2D(img, -1, sharpen_kernel)  
  
plt.imshow(sharpened_img, cmap='gray')  
plt.title("Sharpened Image")  
plt.axis("off")
```

Out[58]: (np.float64(-0.5), np.float64(2930.5), np.float64(1953.5), np.float64(-0.5))

Sharpened Image



```
In [59]: plt.figure(figsize=(10,4))

plt.subplot(1,2,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")

plt.subplot(1,2,2)
plt.imshow(sharpened_img, cmap='gray')
plt.title("Sharpened Image")
plt.axis("off")

plt.show()
```

Original Image



Sharpened Image



Apply sharpening filter on the noisy image.

```
In [61]: import numpy as np
import cv2
import matplotlib.pyplot as plt

sharpen_kernel = np.array([[ 0, -1,  0],
                          [-1,  5, -1],
                          [ 0, -1,  0]])
sharpened_noisy_img = cv2.filter2D(noisy_img, -1, sharpen_kernel)
plt.figure(figsize=(10,4))

plt.subplot(1,2,1)
plt.imshow(noisy_img, cmap='gray')
plt.title("Noisy Image")
plt.axis("off")

plt.subplot(1,2,2)
plt.imshow(sharpened_noisy_img, cmap='gray')
plt.title("Sharpened Noisy Image")
plt.axis("off")

plt.show()
```

Noisy Image



Sharpened Noisy Image



Display original and sharpened images.

```
In [62]: import matplotlib.pyplot as plt

plt.figure(figsize=(8,4))

plt.subplot(1,2,1)
plt.imshow(img, cmap='gray')
plt.title("Original Image")
plt.axis("off")
```

```
plt.subplot(1,2,2)
plt.imshow(sharpened_img, cmap='gray')
plt.title("Sharpened Image")
plt.axis("off")

plt.show()
```

Original Image



Sharpened Image



Mean Squared Error (MSE)

```
In [63]: import numpy as np

mse = np.mean((img.astype(np.float32) - sharpened_img.astype(np.float32)) ** 2)

print("Mean Squared Error (MSE):", mse)
```

Mean Squared Error (MSE): 340.6078

Peak Signal-to-Noise Ratio (PSNR)

```
In [64]: import math

if mse == 0:
    psnr = float('inf')
else:
    max_pixel = 255.0
    psnr = 20 * math.log10(max_pixel / math.sqrt(mse))

print("Peak Signal-to-Noise Ratio (PSNR):", psnr, "dB")
```

Peak Signal-to-Noise Ratio (PSNR): 22.808257868711635 dB