1. Read image, Display it using imshow and matplotlib.

import cv2

```
# Read the image
image_path = 'C:/Users/rcc/Downloads/pexels.jpg'
img = cv2.imread(image_path)
if img is None:
    print("Error: Image not found. Check the path.")
else:
    # Convert to RGB
    img_rgb = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)
    # Display
    plt.imshow(img_rgb)
    plt.title("Original Image")
    plt.axis("off")
    plt.show()
```

Original Image



Perform functions:

• Implement multiplication and division by constant function.

import numpy as np

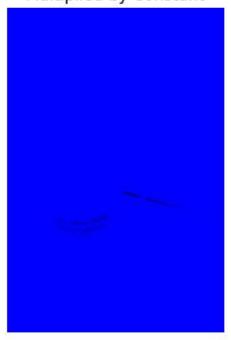
```
# Multiplication and division
img_mul = cv2.multiply(img, np.array([2.0])) # Brighten image
img_div = cv2.divide(img, np.array([2.0])) # Dim image
```

```
# Display results
plt.subplot(1, 2, 1)
plt.imshow(cv2.cvtColor(img_mul, cv2.COLOR_BGR2RGB))
plt.title("Multiplied by Constant")
plt.axis("off")

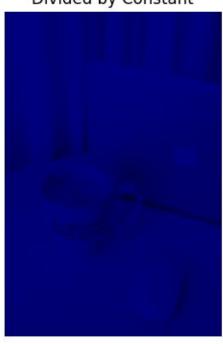
plt.subplot(1, 2, 2)
plt.imshow(cv2.cvtColor(img_div, cv2.COLOR_BGR2RGB))
plt.title("Divided by Constant")
plt.axis("off")
```

plt.show()

Multiplied by Constant



Divided by Constant



• Resize image

```
# Resize the image
img_resized = cv2.resize(img, (200, 200))
```

```
# Display the resized image
plt.imshow(cv2.cvtColor(img_resized, cv2.COLOR_BGR2RGB))
plt.title("Resized Image (200x200)")
plt.axis("off")
plt.show()
```

Resized Image (200x200)



• Slice part of image

Slice a part of the image (e.g., top-left corner) img_sliced = img[:100, :100]

Display the sliced part
plt.imshow(cv2.cvtColor(img_sliced, cv2.COLOR_BGR2RGB))
plt.title("Sliced Image (Top-left 100x100)")
plt.axis("off")
plt.show()

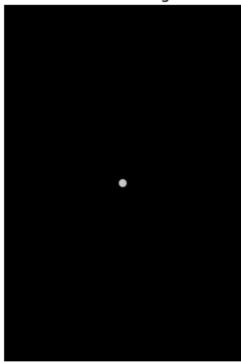
Sliced Image (Top-left 100x100)

Mask a part of image

```
mask = np.zeros_like(img, dtype=np.uint8)
center = (img.shape[1] // 2, img.shape[0] // 2)
radius = 50
cv2.circle(mask, center, radius, (255, 255, 255), -1)
img_masked = cv2.bitwise_and(img, mask)

plt.imshow(cv2.cvtColor(img_masked, cv2.COLOR_BGR2RGB))
plt.title("Masked Image")
plt.axis("off")
plt.show()
```

Masked Image



Add two images

```
# Adding the image to itself
img_add = cv2.add(img, img)

plt.imshow(cv2.cvtColor(img_add, cv2.COLOR_BGR2RGB))
plt.title("Added Image")
plt.axis("off")
plt.show()
```



Weighted add two SWimages

Weighted addition
img_weighted = cv2.addWeighted(img, 0.7, img, 0.3, 0)

plt.imshow(cv2.cvtColor(img_weighted, cv2.COLOR_BGR2RGB))
plt.title("Weighted Added Image")
plt.axis("off")
plt.show()

Weighted Added Image

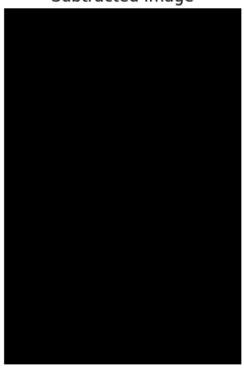


Subtract two image

Subtract the image from itself
img_subtract = cv2.subtract(img, img)

```
plt.imshow(cv2.cvtColor(img_subtract, cv2.COLOR_BGR2RGB))
plt.title("Subtracted Image")
plt.axis("off")
plt.show()
```

Subtracted Image



• Perform logical AND, OR, NOT on image.

```
# Logical AND, OR, NOT
img and = cv2.bitwise and(img, img)
img_or = cv2.bitwise_or(img, img)
img_not = cv2.bitwise_not(img)
# Display logical operations
plt.figure(figsize=(10, 5))
plt.subplot(1, 3, 1)
plt.imshow(cv2.cvtColor(img_and, cv2.COLOR_BGR2RGB))
plt.title("Logical AND")
plt.axis("off")
plt.subplot(1, 3, 2)
plt.imshow(cv2.cvtColor(img or, cv2.COLOR BGR2RGB))
plt.title("Logical OR")
plt.axis("off")
plt.subplot(1, 3, 3)
plt.imshow(cv2.cvtColor(img_not, cv2.COLOR_BGR2RGB))
```

```
plt.title("Logical NOT")
plt.axis("off")
```

plt.show()

Logical AND



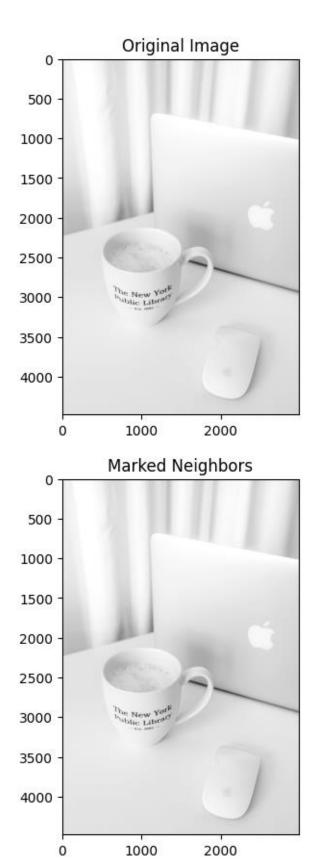




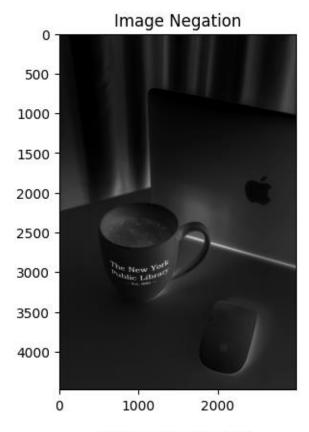
Mark four neighbors and eight neighbors of any pixel in the image, implement distance formula, implement image negation, log transformation, and power log Transformation.

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
image = cv2.imread('C:/Users/rc/Downloads/pexels.jpg', cv2.IMREAD GRAYSCALE)
plt.imshow(image, cmap='gray')
plt.title("Original Image")
plt.show()
# 1. Mark four neighbors and eight neighbors of a pixel
def mark_neighbors(image, x, y):
  marked image = image.copy()
  rows, cols = image.shape
  # Ensure x, y are within bounds
  if x < 0 or x \ge rows or y < 0 or y \ge rows
    raise ValueError("Coordinates out of bounds")
  # Four neighbors
  neighbors_4 = [(x-1, y), (x+1, y), (x, y-1), (x, y+1)]
  # Eight neighbors
  neighbors 8 = \text{neighbors } 4 + [(x-1, y-1), (x-1, y+1), (x+1, y-1), (x+1, y+1)]
  for nx, ny in neighbors_8:
    if 0 \le nx \le ny \le cols:
      marked image[nx, ny] = 255 # Highlight neighbors in white
```

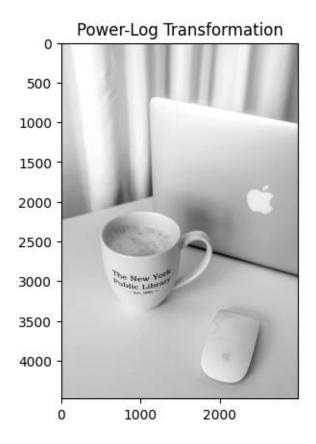
```
return marked image
x, y = 50, 50 \# Example pixel
marked neighbors = mark neighbors(image, x, y)
plt.imshow(marked_neighbors, cmap='gray')
plt.title("Marked Neighbors")
plt.show()
# 2. Implementing distance formula
def distance_formula(p1, p2):
  return np.sqrt((p1[0] - p2[0])**2 + (p1[1] - p2[1])**2)
point1 = (x, y)
point2 = (70, 70)
distance = distance formula(point1, point2)
print(f"Distance between {point1} and {point2}: {distance}")
#3. Image Negation
image negation = 255 - image
plt.imshow(image_negation, cmap='gray')
plt.title("Image Negation")
plt.show()
# 4. Log Transformation
def log transform(image, c=1):
  return (c * np.log1p(image)).astype(np.uint8)
log transformed = log transform(image, c=45)
plt.imshow(log_transformed, cmap='gray')
plt.title("Log Transformation")
plt.show()
# 5. Power-Log (Gamma) Transformation
def power_log_transform(image, gamma, c=1):
  normalized = image / 255.0 # Normalize to range [0,1]
  transformed = c * (normalized ** gamma)
  return (transformed * 255).astype(np.uint8)
gamma = 2.2
power log transformed = power log transform(image, gamma, c=1)
plt.imshow(power log transformed, cmap='gray')
plt.title("Power-Log Transformation")
plt.show()
```



Distance between (50, 50) and (70, 70): 28.284271247461902







Gray level Slicing, Intensity slicing, Boxplot slicing, Bit plane slicing, Histogram Equalizer

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

Load the grayscale image image = cv2.imread('C:/Users/rcc/Downloads/pexels.jpg', cv2.IMREAD_GRAYSCALE)

Display original image plt.imshow(image, cmap='gray') plt.title("Original Image") plt.show()

1. Gray Level Slicing

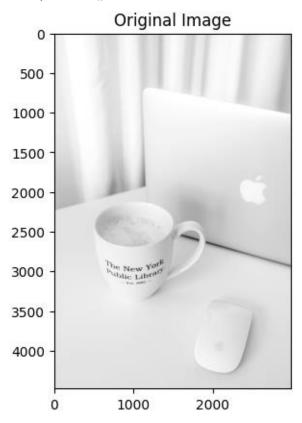
def gray_level_slicing(image, lower, upper, highlight_value=255, background_value=0):
 sliced_image = np.where((image >= lower) & (image <= upper), highlight_value, background_value)
 return sliced_image.astype(np.uint8)</pre>

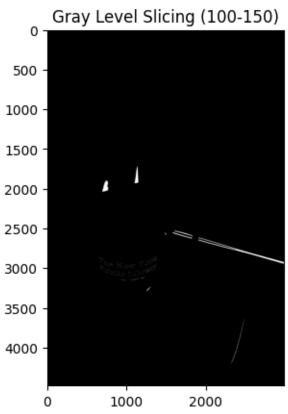
gray_sliced = gray_level_slicing(image, lower=100, upper=150)
plt.imshow(gray_sliced, cmap='gray')
plt.title("Gray Level Slicing (100-150)")
plt.show()

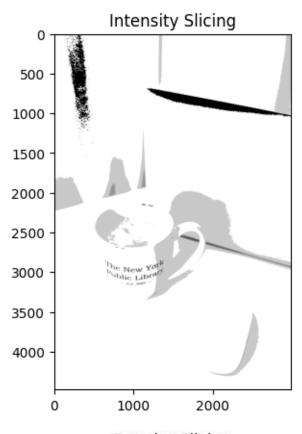
2. Intensity Slicing def intensity_slicing(image, levels): sliced_image = np.zeros_like(image) interval = 256 // len(levels)

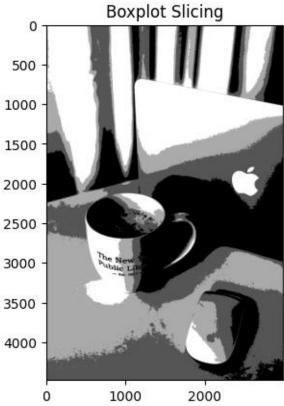
```
for i, value in enumerate(levels):
    lower = i * interval
    upper = lower + interval
    sliced image[(image >= lower) & (image < upper)] = value
  return sliced_image
levels = [50, 100, 150, 200, 255] # Example intensity levels
intensity sliced = intensity slicing(image, levels)
plt.imshow(intensity sliced, cmap='gray')
plt.title("Intensity Slicing")
plt.show()
#3. Boxplot Slicing
def boxplot slicing(image):
  boxplot values = np.percentile(image, [25, 50, 75]) # Calculate Q1, Median, Q3
  sliced_image = np.where(image < boxplot_values[0], 0,
           np.where(image < boxplot_values[1], 85,
           np.where(image < boxplot values[2], 170, 255)))
  return sliced_image.astype(np.uint8)
boxplot sliced = boxplot slicing(image)
plt.imshow(boxplot sliced, cmap='gray')
plt.title("Boxplot Slicing")
plt.show()
# 4. Bit Plane Slicing
def bit_plane_slicing(image, bit_level):
  return ((image >> bit_level) & 1) * 255
# Displaying all 8 bit planes
bit planes = [bit plane slicing(image, i) for i in range(8)]
fig, axes = plt.subplots(2, 4, figsize=(15, 6))
for i, ax in enumerate(axes.flatten()):
  ax.imshow(bit planes[i], cmap='gray')
  ax.set_title(f'Bit Plane {i}')
  ax.axis('off')
plt.tight layout()
plt.show()
# 5. Histogram Equalization
equalized image = cv2.equalizeHist(image)
plt.imshow(equalized image, cmap='gray')
plt.title("Histogram Equalization")
plt.show()
plt.figure(figsize=(10, 5))
plt.hist(image.ravel(), bins=256, range=(0, 256), alpha=0.5, label='Original')
plt.hist(equalized_image.ravel(), bins=256, range=(0, 256), alpha=0.5, label='Equalized')
```

plt.title("Histogram Comparison")
plt.legend()
plt.show()



















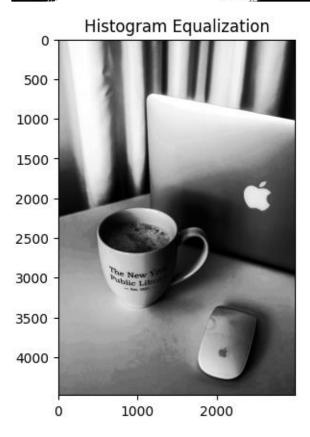




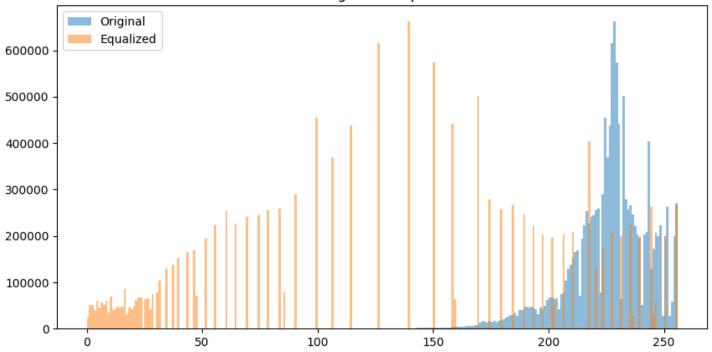








Histogram Comparison



import urllib.request

url = 'https://raw.githubusercontent.com/opencv/opencv/master/samples/data/lena.jpg'
urllib.request.urlretrieve(url, 'lena.jpg')

('lena.jpg', <http.client.HTTPMessage at 0x17b016063c0>)

Box filters, filter 2D, Gaussian Blur, and Median Blur, & write a program to hardcode convolution using a kernel of known size & value

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

Load the image image = cv2.imread('lena.jpg', cv2.IMREAD_GRAYSCALE)

Display original image plt.imshow(image, cmap='gray') plt.title("Original Image") plt.axis('off') plt.show()

1. Box Filter box_filter = cv2.blur(image, (5, 5)) # Kernel size: 5x5 plt.imshow(box_filter, cmap='gray')

```
plt.title("Box Filter (5x5)")
plt.axis('off')
plt.show()
#2. Filter 2D
kernel = np.array([[1, 1, 1],
          [1, 1, 1],
          [1, 1, 1]], dtype=np.float32) / 9 # 3x3 averaging kernel
filter_2d = cv2.filter2D(image, -1, kernel)
plt.imshow(filter 2d, cmap='gray')
plt.title("2D Filter (3x3 Averaging)")
plt.axis('off')
plt.show()
#3. Gaussian Blur
gaussian blur = cv2.GaussianBlur(image, (5, 5), sigmaX=1)
plt.imshow(gaussian_blur, cmap='gray')
plt.title("Gaussian Blur (5x5, sigma=1)")
plt.axis('off')
plt.show()
#4. Median Blur
median blur = cv2.medianBlur(image, 5) # Kernel size: 5
plt.imshow(median blur, cmap='gray')
plt.title("Median Blur (5x5)")
plt.axis('off')
plt.show()
# 5. Hardcoded Convolution
def hardcoded_convolution(image, kernel):
  kernel height, kernel width = kernel.shape
  pad_h, pad_w = kernel_height // 2, kernel_width // 2
  # Pad the image with zeros
  padded_image = np.pad(image, ((pad_h, pad_h), (pad_w, pad_w)), mode='constant',
constant values=0)
  # Prepare output image
  output = np.zeros like(image)
  # Perform convolution
  for i in range(image.shape[0]):
```

Original Image

for j in range(image.shape[1]):



Box Filter (5x5)



2D Filter (3x3 Averaging)



Gaussian Blur (5x5, sigma=1)



Median Blur (5x5)



Custom Convolution (Sharpening)



Plot the histogram after and before the effects (whatever effects are given above)

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
# Load the grayscale image
image = cv2.imread('lena.jpg', cv2.IMREAD_GRAYSCALE)
# Display original image
plt.imshow(image, cmap='gray')
plt.title("Original Image")
plt.axis('off')
plt.show()
# Function to plot histogram
def plot_histogram(image, title):
  plt.hist(image.ravel(), bins=256, range=(0, 256), color='gray', alpha=0.7)
  plt.title(title)
  plt.xlabel("Pixel Intensity")
  plt.ylabel("Frequency")
  plt.show()
# Plot original histogram
plot_histogram(image, "Histogram (Original Image)")
#1. Box Filter
box_filter = cv2.blur(image, (5, 5))
plot_histogram(box_filter, "Histogram (Box Filter)")
```

```
#2.2D Filter
kernel = np.array([[1, 1, 1],
          [1, 1, 1],
          [1, 1, 1]], dtype=np.float32) / 9
filter_2d = cv2.filter2D(image, -1, kernel)
plot histogram(filter 2d, "Histogram (2D Filter)")
#3. Gaussian Blur
gaussian blur = cv2.GaussianBlur(image, (5, 5), sigmaX=1)
plot_histogram(gaussian_blur, "Histogram (Gaussian Blur)")
#4. Median Blur
median blur = cv2.medianBlur(image, 5)
plot_histogram(median_blur, "Histogram (Median Blur)")
# 5. Hardcoded Convolution
def hardcoded_convolution(image, kernel):
  kernel height, kernel width = kernel.shape
  pad_h, pad_w = kernel_height // 2, kernel_width // 2
 # Pad the image with zeros
  padded image = np.pad(image, ((pad h, pad h), (pad w, pad w)), mode='constant', constant values=0)
 # Prepare output image
  output = np.zeros_like(image)
 # Perform convolution
 for i in range(image.shape[0]):
    for j in range(image.shape[1]):
      region = padded image[i:i+kernel height, j:j+kernel width]
      output[i, j] = np.sum(region * kernel)
 # Normalize output to range 0-255
  output = np.clip(output, 0, 255).astype(np.uint8)
  return output
# Example kernel for convolution
custom_kernel = np.array([[0, -1, 0],
              [-1, 5, -1],
              [0, -1, 0]]) # Example sharpening kernel
custom convolution = hardcoded convolution(image, custom kernel)
plot histogram(custom convolution, "Histogram (Custom Convolution)")
# Display all processed images
effects = {
  "Box Filter": box filter,
  "2D Filter": filter 2d,
  "Gaussian Blur": gaussian_blur,
  "Median Blur": median_blur,
```

```
"Custom Convolution": custom_convolution,
}

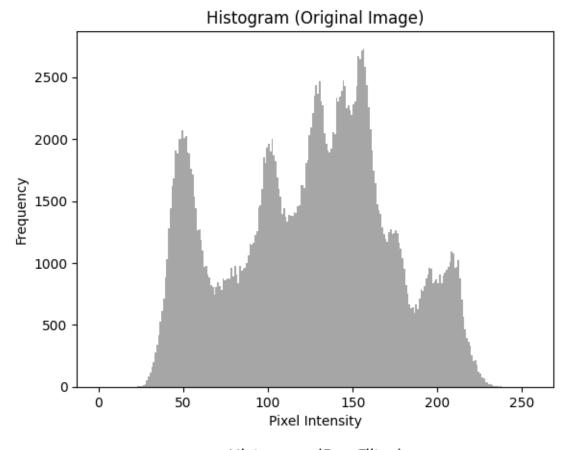
fig, axes = plt.subplots(2, 3, figsize=(15, 10))
axes = axes.ravel()

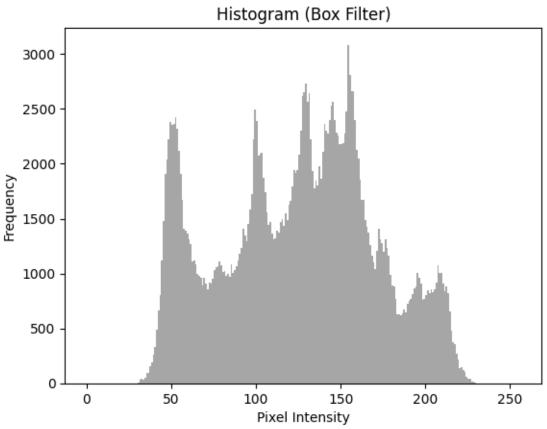
# Show processed images
for idx, (title, img) in enumerate(effects.items()):
    axes[idx].imshow(img, cmap='gray')
    axes[idx].set_title(title)
    axes[idx].axis('off')

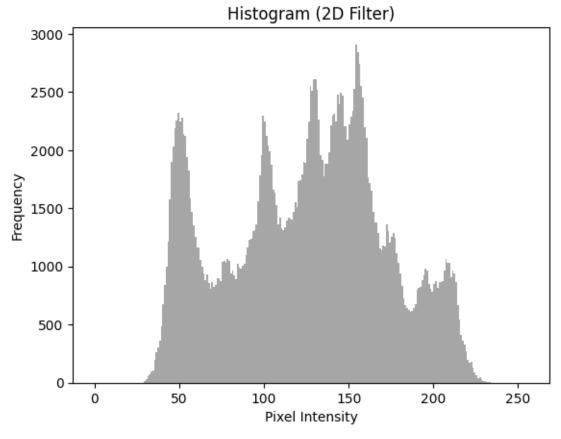
plt.tight_layout()
plt.show()
```

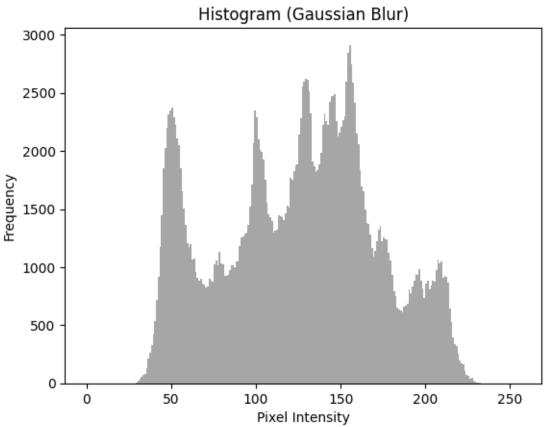
Original Image

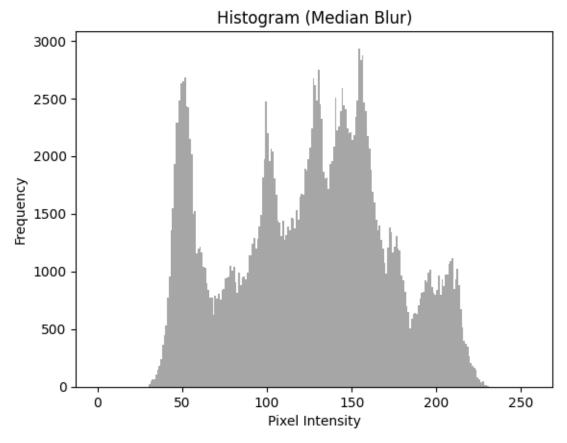


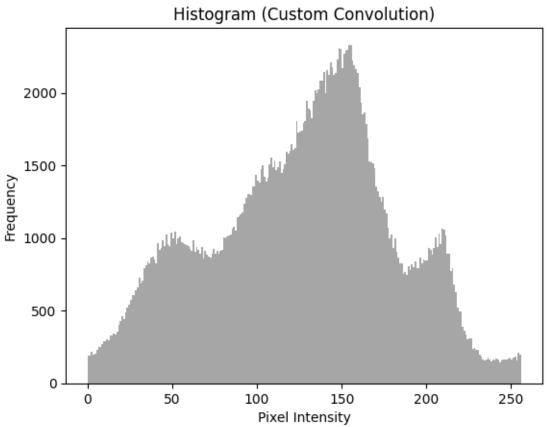


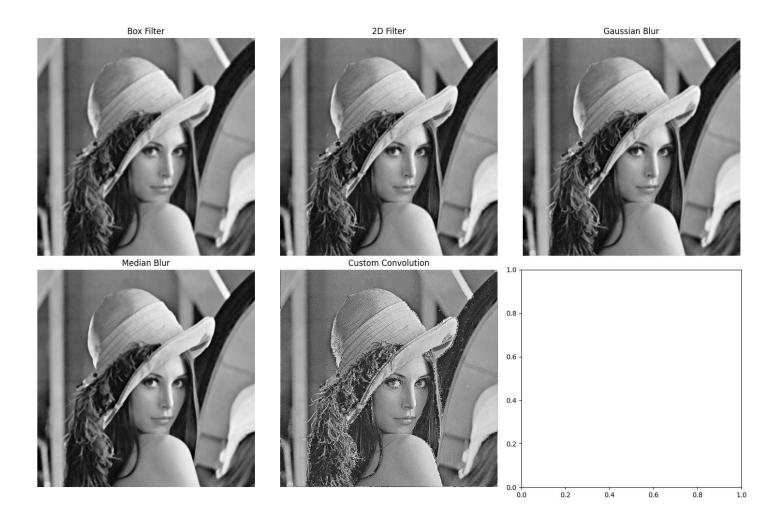












Edge detection using Roberts, Sobel, Prewitt, and Kenny using 2D filters as well as openCV functions.

```
import cv2
import matplotlib.pyplot as plt

# Define the edge detection functions

def roberts_edge_detection(image):
    image = np.float32(image)
    kernel_x = np.array([[1, 0], [0, -1]])
    kernel_y = np.array([[0, 1], [-1, 0]])

grad_x = cv2.filter2D(image, -1, kernel_x)
    grad_y = cv2.filter2D(image, -1, kernel_y)
    grad = cv2.sqrt(grad_x**2 + grad_y**2) # Combine gradients
    return grad

def sobel_edge_detection(image):
    image = np.float32(image) # Convert to float32 for calculations
    kernel_x = np.array([[-1, 0, 1], [-2, 0, 2], [-1, 0, 1]])
    kernel_y = np.array([[-1, -2, -1], [0, 0, 0], [1, 2, 1]])
```

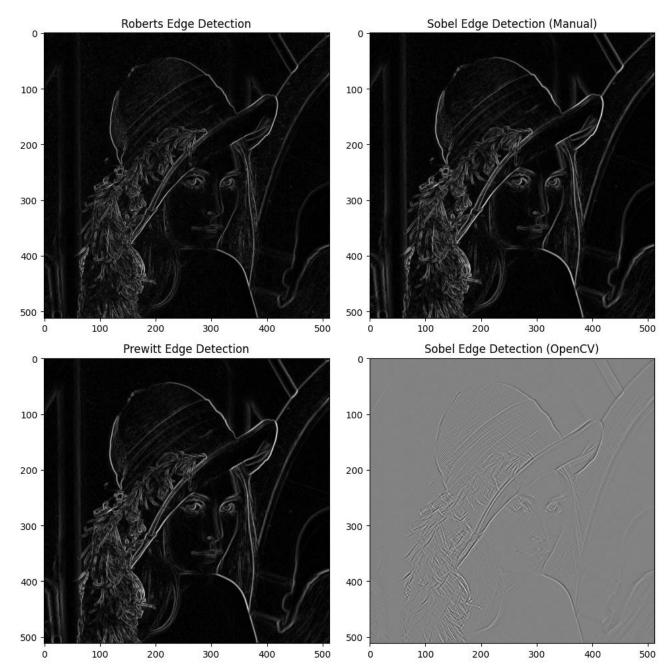
grad_x = cv2.filter2D(image, -1, kernel_x)
grad_y = cv2.filter2D(image, -1, kernel_y)

import numpy as np

```
grad = cv2.sqrt(grad x^{**}2 + grad y^{**}2) # Combine gradients
  return grad
def prewitt edge detection(image):
  image = np.float32(image) # Convert to float32 for calculations
  kernel_x = np.array([[-1, 0, 1], [-1, 0, 1], [-1, 0, 1]])
  kernel_y = np.array([[-1, -1, -1], [0, 0, 0], [1, 1, 1]])
  grad x = cv2.filter2D(image, -1, kernel x)
  grad y = cv2.filter2D(image, -1, kernel y)
  grad = cv2.sqrt(grad x**2 + grad y**2) # Combine gradients
  return grad
# Load the image in grayscale
image = cv2.imread('lena.jpg', cv2.IMREAD GRAYSCALE)
edges_roberts = roberts_edge_detection(image)
edges sobel = sobel edge detection(image)
edges prewitt = prewitt edge detection(image)
edges sobel opencv = cv2.Sobel(image, cv2.CV 64F, 1, 1, ksize=3)
# Normalize
edges roberts normalized = cv2.normalize(edges roberts, None, 0, 255, cv2.NORM MINMAX)
edges roberts normalized = np.uint8(edges roberts normalized)
edges_sobel_normalized = cv2.normalize(edges_sobel, None, 0, 255, cv2.NORM_MINMAX)
edges_sobel_normalized = np.uint8(edges_sobel_normalized)
edges prewitt normalized = cv2.normalize(edges prewitt, None, 0, 255, cv2.NORM MINMAX)
edges prewitt normalized = np.uint8(edges prewitt normalized)
edges sobel opency normalized = cv2.normalize(edges sobel opency, None, 0, 255,
cv2.NORM MINMAX)
edges sobel_opencv_normalized = np.uint8(edges_sobel_opencv_normalized)
plt.figure(figsize=(10, 10))
plt.subplot(2, 2, 1)
plt.imshow(edges_roberts_normalized, cmap='gray')
plt.title('Roberts Edge Detection')
plt.subplot(2, 2, 2)
plt.imshow(edges sobel normalized, cmap='gray')
plt.title('Sobel Edge Detection (Manual)')
plt.subplot(2, 2, 3)
plt.imshow(edges prewitt normalized, cmap='gray')
plt.title('Prewitt Edge Detection')
```

```
plt.subplot(2, 2, 4)
plt.imshow(edges_sobel_opencv_normalized, cmap='gray')
plt.title('Sobel Edge Detection (OpenCV)')

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



Perform transformations on image using hardcoding with matrix multiplication of image and matrices of transformation

```
import numpy as np
import cv2
import matplotlib.pyplot as plt
def transform_image(image, transformation_matrix):
   rows, cols = image.shape
   transformed_image = np.zeros_like(image)
   for i in range(rows):
```

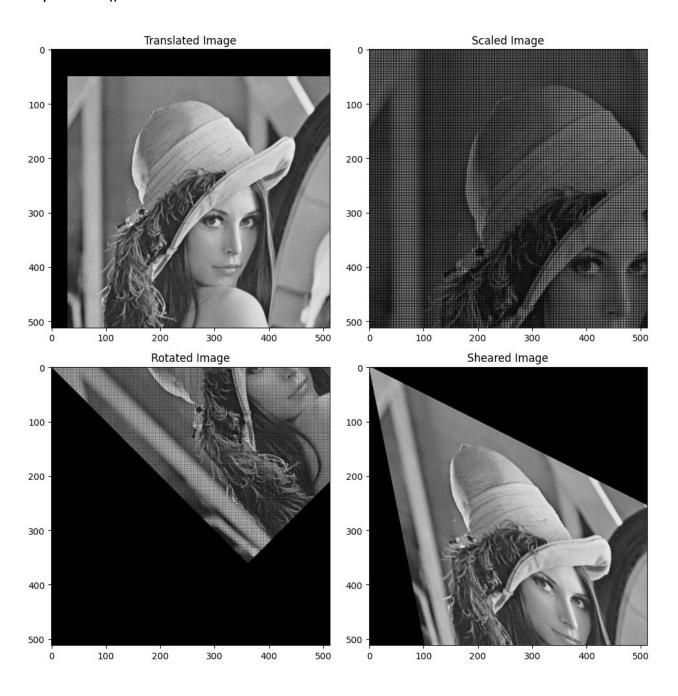
```
for j in range(cols):
       original coords = np.array([i, j, 1])
      transformed coords = np.dot(transformation matrix, original coords)
       new_x, new_y = int(transformed_coords[0]), int(transformed_coords[0])
       if 0 \le \text{new } x < \text{cols and } 0 \le \text{new } y < \text{rows}:
         transformed_image[new_y, new_x] = image[i, j]
  return transformed image
image = cv2.imread('lena.jpg', cv2.IMREAD_GRAYSCALE)
# Translation matrix (shift image 50 pixels right, 30 pixels down)
T = np.array([[1, 0, 50],
        [0, 1, 30],
        [0, 0, 1]]
# Scaling matrix (scale by 1.5 in x and 1.5 in y)
S = np.array([[1.5, 0, 0],
        [0, 1.5, 0],
        [0, 0, 1]]
# Rotation matrix (rotate by 45 degrees)
theta = 45
theta = np.radians(theta) # Convert to radians
R = np.array([[np.cos(theta), -np.sin(theta), 0],
        [np.sin(theta), np.cos(theta), 0],
        [0, 0, 1]
# Shear matrix (shear by 0.5 in x direction and 0.2 in y direction)
H = np.array([[1, 0.5, 0],
        [0.2, 1, 0],
        [0, 0, 1]
# Apply transformations
image translated = transform image(image, T)
image scaled = transform image(image, S)
image rotated = transform_image(image, R)
image sheared = transform image(image, H)
plt.figure(figsize=(10, 10))
plt.subplot(2, 2, 1)
plt.imshow(image translated, cmap='gray')
plt.title('Translated Image')
```

plt.subplot(2, 2, 2)
plt.imshow(image_scaled, cmap='gray')
plt.title('Scaled Image')

plt.subplot(2, 2, 3)
plt.imshow(image_rotated, cmap='gray')
plt.title('Rotated Image')

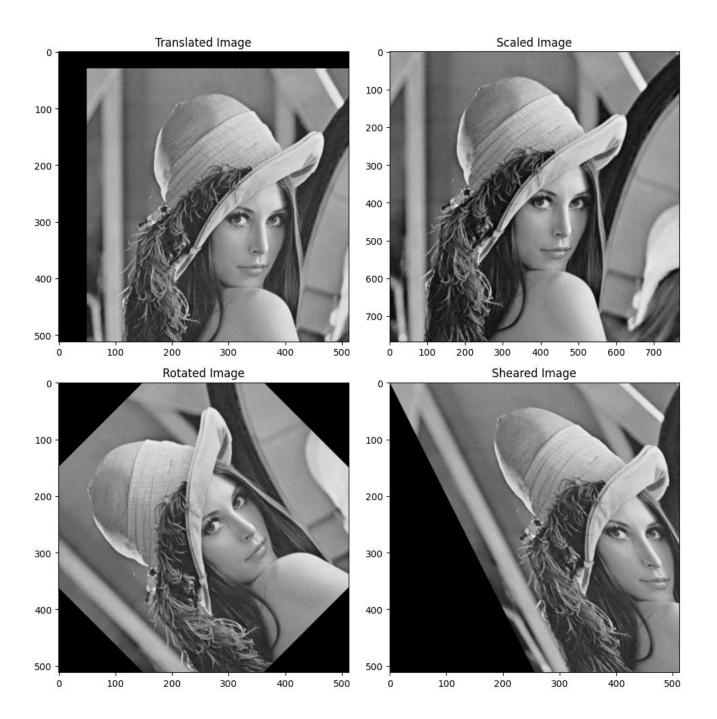
plt.subplot(2, 2, 4)
plt.imshow(image_sheared, cmap='gray')
plt.title('Sheared Image')

plt.tight_layout()
plt.show()



Perform transformation using the OpenCV library

```
import numpy as np
import cv2
import matplotlib.pyplot as plt
def transform image opency(image, transformation matrix, output size):
  transformed_image = cv2.warpAffine(image, transformation_matrix, output_size)
  return transformed image
image = cv2.imread('image.jpg', cv2.IMREAD_GRAYSCALE)
translation matrix = np.float32([[1, 0, 50], [0, 1, 30]])
height, width = image.shape
image_translated = transform_image_opencv(image, translation_matrix, (width, height))
image_scaled = cv2.resize(image, None, fx=1.5, fy=1.5, interpolation=cv2.INTER_LINEAR)
rotation matrix = cv2.getRotationMatrix2D((width//2, height//2), 45, 1)
image_rotated = transform_image_opencv(image, rotation_matrix, (width, height))
shear_matrix = np.float32([[1, 0.5, 0], [0, 1, 0]])
image_sheared = transform_image_opencv(image, shear_matrix, (width, height))
plt.figure(figsize=(10, 10))
plt.subplot(2, 2, 1)
plt.imshow(image translated, cmap='gray')
plt.title('Translated Image')
plt.subplot(2, 2, 2)
plt.imshow(image_scaled, cmap='gray')
plt.title('Scaled Image')
plt.subplot(2, 2, 3)
plt.imshow(image_rotated, cmap='gray')
plt.title('Rotated Image')
plt.subplot(2, 2, 4)
plt.imshow(image sheared, cmap='gray')
plt.title('Sheared Image')
plt.tight layout()
plt.show()
```



Perform Segmentation using

- Simple thresholding
- Otsu Binarization
- Adaptive Thresholding

import cv2
import numpy as np
import matplotlib.pyplot as plt
image = cv2.imread('lena.jpg', cv2.IMREAD_GRAYSCALE)
_, thresh_simple = cv2.threshold(image, 127, 255, cv2.THRESH_BINARY)
_, thresh_otsu = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
thresh_adaptive = cv2.adaptiveThreshold(image, 255,
cv2.ADAPTIVE_THRESH_GAUSSIAN_C,

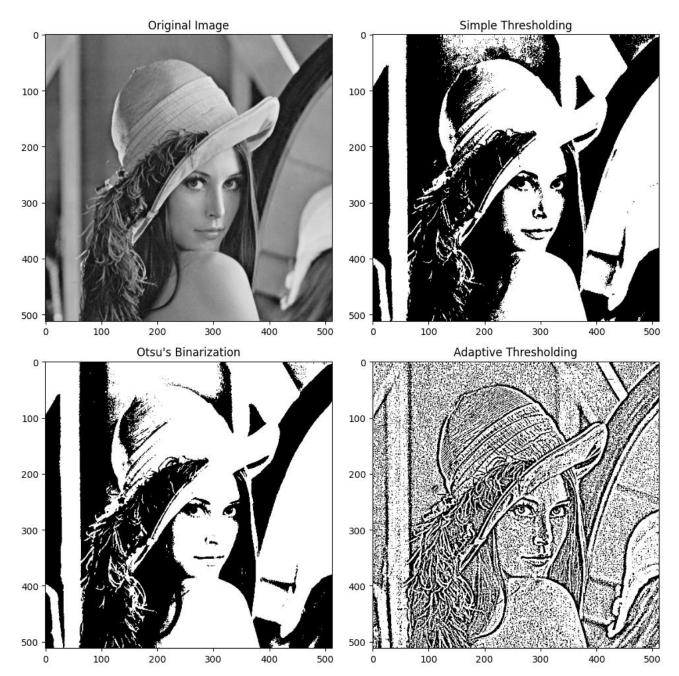
cv2.THRESH_BINARY, 11, 2)

```
plt.figure(figsize=(10, 10))
plt.subplot(2, 2, 1)
plt.imshow(image, cmap='gray')
plt.title('Original Image')

plt.subplot(2, 2, 2)
plt.imshow(thresh_simple, cmap='gray')
plt.title('Simple Thresholding')

plt.subplot(2, 2, 3)
plt.imshow(thresh_otsu, cmap='gray')
plt.title("Otsu's Binarization")

plt.subplot(2, 2, 4)
plt.imshow(thresh_adaptive, cmap='gray')
plt.title('Adaptive Thresholding')
plt.tight_layout()
plt.show()
```



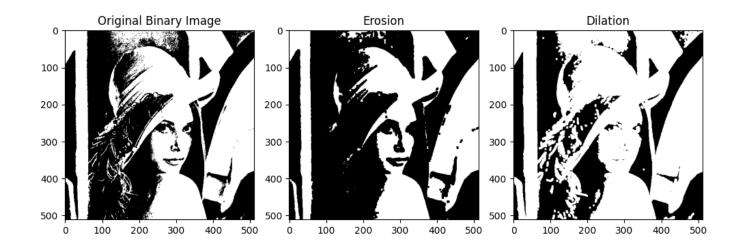
Perform erosion, dilation, opening, closing

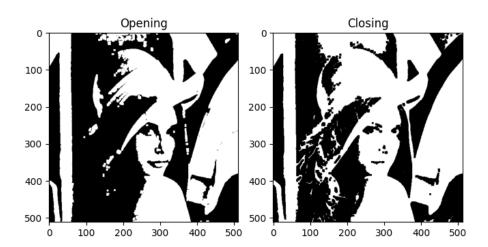
import cv2 import numpy as np import matplotlib.pyplot as plt

image = cv2.imread('lena.jpg', cv2.IMREAD_GRAYSCALE)
_, binary_image = cv2.threshold(image, 127, 255, cv2.THRESH_BINARY)
kernel = np.ones((5, 5), np.uint8)

erosion = cv2.erode(binary_image, kernel, iterations=1)
dilation = cv2.dilate(binary_image, kernel, iterations=1)
opening = cv2.morphologyEx(binary_image, cv2.MORPH_OPEN, kernel)
closing = cv2.morphologyEx(binary_image, cv2.MORPH_CLOSE, kernel)

```
plt.figure(figsize=(10, 10))
plt.subplot(2, 3, 1)
plt.imshow(binary_image, cmap='gray')
plt.title('Original Binary Image')
plt.subplot(2, 3, 2)
plt.imshow(erosion, cmap='gray')
plt.title('Erosion')
plt.subplot(2, 3, 3)
plt.imshow(dilation, cmap='gray')
plt.title('Dilation')
plt.subplot(2, 3, 4)
plt.imshow(opening, cmap='gray')
plt.title('Opening')
plt.subplot(2, 3, 5)
plt.imshow(closing, cmap='gray')
plt.title('Closing')
plt.tight_layout()
plt.show()
```





WAP to implement image compression using K-Means.

import cv2

import numpy as np

import matplotlib.pyplot as plt

```
from sklearn.cluster import KMeans

def compress_image_kmeans(image_path, k=8):
    image = cv2.imread(image_path)
    image_rgb = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
    pixels = image_rgb.reshape(-1, 3)
    kmeans = KMeans(n_clusters=k, random_state=42)
    kmeans.fit(pixels)
    compressed_pixels = kmeans.cluster_centers_[kmeans.labels_]
```

compressed_image = compressed_pixels.reshape(image_rgb.shape).astype(np.uint8)

return compressed_image, kmeans image_path = 'lena.jpg' compressed_image, kmeans = compress_image_kmeans(image_path, k=8)

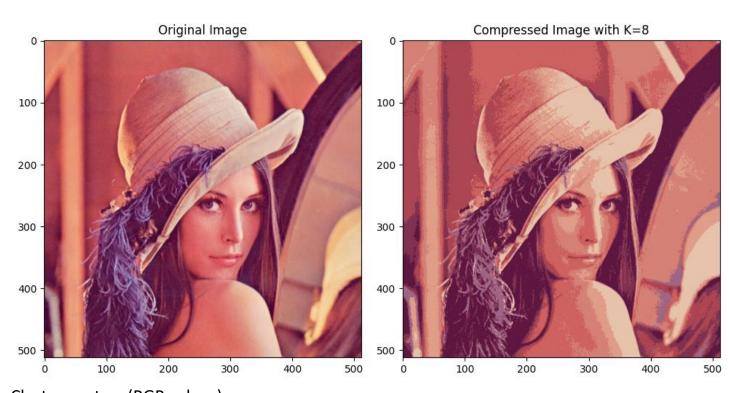
```
image = cv2.imread(image_path)
image_rgb = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)

plt.figure(figsize=(10, 5))

plt.subplot(1, 2, 1)
plt.imshow(image_rgb)
plt.title('Original Image')

plt.subplot(1, 2, 2)
plt.imshow(compressed_image)
plt.title(f'Compressed Image with K={8}')
plt.tight_layout()
plt.show()
```

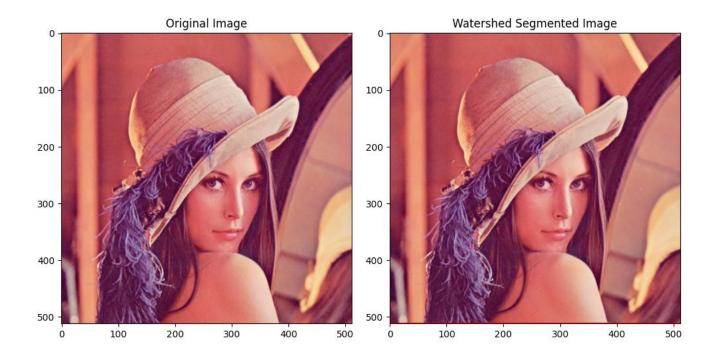
print(kmeans.cluster_centers_)



Cluster centers (RGB values):
[[129.70249267 50.43704616 82.54333517]
[213.16440006 129.2301834 117.67438607]
[172.93762304 70.1508859 82.04239368]
[232.17514473 195.94697028 171.84913161]
[224.3887498 158.81222743 137.56501373]
[93.59140013 23.71865659 64.84226506]
[205.9218581 98.04462676 96.88676524]

WAP to implement Watershed algorithm.

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
def watershed segmentation(image path):
  image = cv2.imread(image_path)
  gray = cv2.cvtColor(image, cv2.COLOR BGR2GRAY)
  , binary = cv2.threshold(gray, 120, 255, cv2.THRESH_BINARY_INV)
  dist_transform = cv2.distanceTransform(binary, cv2.DIST_L2, 5)
  _, markers = cv2.threshold(dist_transform, 0.7 * dist_transform.max(), 255, 0)
  markers = np.int32(markers)
  markers = markers + 1
  markers[binary == 255] = 0
  cv2.watershed(image, markers)
  image[markers == -1] = [0, 0, 255]
  return image, markers
image path = 'image.jpg'
segmented image, markers = watershed segmentation(image path)
image = cv2.imread(image path)
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
plt.imshow(cv2.cvtColor(image, cv2.COLOR BGR2RGB))
plt.title('Original Image')
plt.subplot(1, 2, 2)
plt.imshow(cv2.cvtColor(segmented image, cv2.COLOR BGR2RGB))
plt.title('Watershed Segmented Image')
plt.tight layout()
plt.show()
```



WAP for image classification using traditional ML algorithms.

```
import cv2
import numpy as np
from skimage.feature import hog
from skimage import exposure
from sklearn.svm import SVC
from sklearn.model selection import train test split
from sklearn.metrics import classification_report, accuracy_score
import os
import matplotlib.pyplot as plt
def extract_hog_features(image_path):
  image = cv2.imread("C:/Users/priya/Downloads/dataset/cat/Abyssinian 71.jpg",
cv2.IMREAD GRAYSCALE)
  image_resized = cv2.resize(image, (64, 64))
  features, _ = hog(image_resized, pixels_per_cell=(8, 8), cells_per_block=(2, 2),
visualize=True)
  return features
image directory = 'dataset'
categories = ['cat', 'dog']
features_list = []
labels_list = []
```

Loop through the categories and load images

```
for label, category in enumerate(categories):
  category_folder = os.path.join(image_directory, category)
  for filename in os.listdir(category folder):
    # Full image path
    image_path = os.path.join(category_folder, filename)
    # Extract HOG features from the image
    features = extract_hog_features(image_path)
    # Append features and corresponding label to the lists
    features_list.append(features)
    labels list.append(label)
# Convert lists to numpy arrays
X = np.array(features list) # Feature matrix
y = np.array(labels_list) # Label vector
# Split data into training and testing sets (80% training, 20% testing)
X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
# Initialize the SVM classifier (Support Vector Machine)
svm classifier = SVC(kernel='linear', random state=42)
# Train the classifier using the training data
svm classifier.fit(X train, y train)
# Predict on the test set
y pred = svm classifier.predict(X test)
# Evaluate the classifier
print("Confusion Matrix:")
print(confusion matrix(y test, y pred))
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Classification Report:")
print(classification report(y test, y pred, target names=categories))
# Optionally, visualize some test images and predictions
fig, axes = plt.subplots(1, 5, figsize=(15, 5))
for i in range(5):
  ax = axes[i]
```

```
img_path = os.path.join(image_directory, categories[y_test[i]],
os.listdir(os.path.join(image_directory, categories[y_test[i]]))[i])
img = cv2.imread("C:/Users/priya/Downloads/dataset/cat/Abyssinian_129.jpg")
img_rgb = cv2.cvtColor(img, cv2.COLOR_BGR2RGB)

ax.imshow(img_rgb)
ax.set_title(f"True: {categories[y_test[i]]}\nPred: {categories[y_pred[i]]}")
ax.axis('off')

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

Confusion Matrix:

[[0 46]

[0 35]]

Accuracy: 0.43209876543209874

Classification Report:

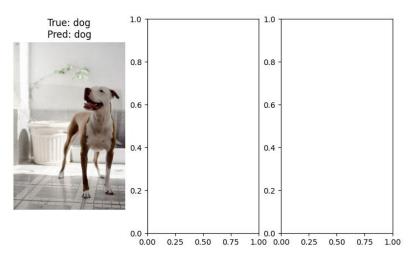
precision recall f1-score support

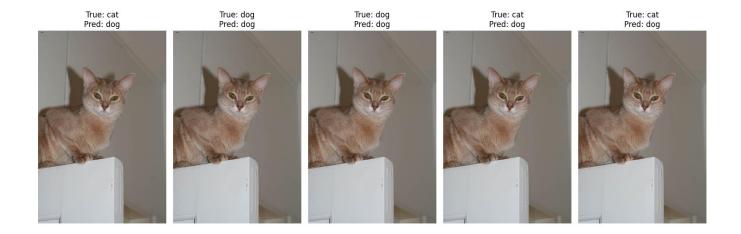
cat 0.00 0.00 0.00 46 dog 0.43 1.00 0.60 35

accuracy 0.43 81 macro avg 0.22 0.50 0.30 81 weighted avg 0.19 0.43 0.26 81









WAP to extract Haris corner detection feature.

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
def harris_corner_detection(image_path):
  image = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
  image float = np.float32(image)
  corners = cv2.cornerHarris(image_float, blockSize=2, ksize=3, k=0.04)
  corners_dilated = cv2.dilate(corners, None)
  image with corners = np.copy(image)
  image_with_corners[corners_dilated > 0.01 * corners_dilated.max()] = 255
  return image_with_corners, corners_dilated
image_path = 'lena.jpg'
image_with_corners, corners_dilated = harris_corner_detection(image_path)
image color = cv2.imread(image path)
image color rgb = cv2.cvtColor(image color, cv2.COLOR BGR2RGB)
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
plt.imshow(image color rgb)
plt.title('Original Image')
plt.axis('off')
plt.subplot(1, 2, 2)
plt.imshow(image_with_corners, cmap='gray')
plt.title('Harris Corners Detected')
plt.axis('off')
plt.tight_layout()
plt.show()
```

Original Image





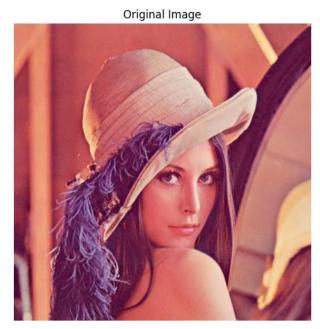


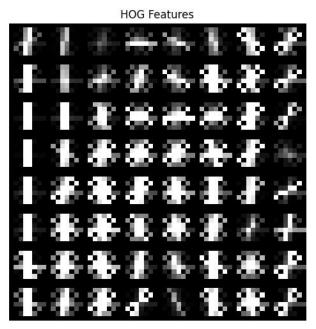
WAP to extract HOG features.

import cv2
import numpy as np
from skimage.feature import hog
from skimage import exposure
import matplotlib.pyplot as plt

```
def extract_hog_features(image_path):
  image = cv2.imread(image path)
  gray image = cv2.cvtColor(image, cv2.COLOR BGR2GRAY)
  gray_image_resized = cv2.resize(gray_image, (64, 64))
  features, hog_image = hog(gray_image_resized, pixels_per_cell=(8, 8),
cells_per_block=(2, 2), visualize=True)
  hog_image_rescaled = exposure.rescale_intensity(hog_image, in_range=(0, 10))
  return features, hog_image_rescaled
image_path = 'lena.jpg'
features, hog_image_rescaled = extract_hog_features(image_path)
original image = cv2.imread(image_path)
original image rgb = cv2.cvtColor(original image, cv2.COLOR BGR2RGB)
plt.figure(figsize=(10, 5))
plt.subplot(1, 2, 1)
plt.imshow(original_image_rgb)
plt.title('Original Image')
plt.axis('off')
plt.subplot(1, 2, 2)
```

```
plt.imshow(hog_image_rescaled, cmap='gray')
plt.title('HOG Features')
plt.axis('off')
plt.tight_layout()
plt.show()
print(f"Length of HOG feature vector: {len(features)}")
```





Length of HOG feature vector: 1764

WAP to extract SIFT features.

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

def extract_sift_features(image_path):
    image = cv2.imread(image_path)
    gray_image = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
    sift = cv2.SIFT_create()
    keypoints, descriptors = sift.detectAndCompute(gray_image, None)
    return image, keypoints, descriptors

image_path = 'lena.jpg'
image_with_keypoints, keypoints, descriptors = extract_sift_features(image_path)
image_rgb = cv2.cvtColor(image_with_keypoints, cv2.COLOR_BGR2RGB)
image_with_keypoints = cv2.drawKeypoints(image_rgb, keypoints, None,
```

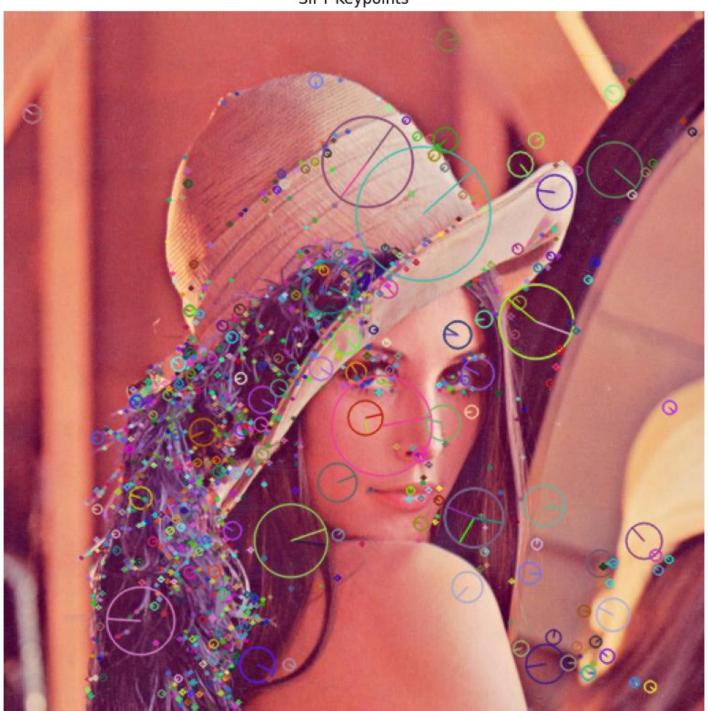
flags=cv2.DRAW MATCHES FLAGS DRAW RICH KEYPOINTS)

```
plt.figure(figsize=(10, 10))
plt.imshow(image_with_keypoints)
```

```
plt.title('SIFT Keypoints')
plt.axis('off')
plt.show()
```

print(f"Number of keypoints detected: {len(keypoints)}")





Number of keypoints detected: 1111