



CSE438

Section: 01

Lab: 05 Report

Topic: Unsharp Masking, High Boost Filtering, Edge Detection Submitted By:

Name: Md Sifat Ullah Sheikh

ID: 2022-1-60-029

Submitted To:

K.M. Safin Kamal

Lecturer

Department of Computer Science & Engineering

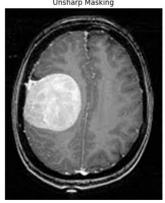
Date: 8 August 2025

- Q1. Sharpen the following image by applying the following and find out which one is better:
- a) Unsharp Masking
- b) High Boost Filtering

CODE:

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
img = cv2.imread('/kaggle/input/lab-05-dataset/lab 05/Picture1.jpg')
img = cv2.cvtColor(img, cv2.COLOR BGR2RGB)
blurred = cv2.GaussianBlur(img, (5, 5), 0)
unsharp mask = cv2.addWeighted(src1=img, alpha=1.5, src2=blurred, beta=-0.5, gamma=0)
A = 2.0
high boost = cv2.addWeighted(src1=img, alpha=A, src2=blurred, beta=-(A - 1), gamma=0)
titles = ['Original Image', 'Unsharp Masking', 'High Boost Filtering']
images = [img, unsharp mask, high boost]
plt.figure(figsize=(15, 5))
for i in range(3):
  plt.subplot(1, 3, i + 1)
  plt.imshow(images[i])
  plt.title(titles[i])
  plt.axis('off')
plt.tight layout()
plt.show()
                                           Unsharp Masking
```



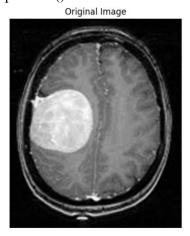




- After Tunning the parameter K value ------

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
# Read the image
img = cv2.imread('/kaggle/input/lab-05-dataset/lab 05/Picture1.jpg')
img = cv2.cvtColor(img, cv2.COLOR BGR2RGB)
# Unsharp Masking
blurred = cv2.GaussianBlur(img, (0, 0), 3)
unsharp mask = cv2.addWeighted(img, 1.5, blurred, -0.5, 0)
# High Boost Filtering (k=2)
high boost = cv2.addWeighted(img, 3, blurred, -1.5, 0)
# Display results
plt.figure(figsize=(15, 10))
plt.subplot(1, 3, 1)
plt.imshow(img)
plt.title('Original Image')
plt.axis('off')
plt.subplot(1, 3, 2)
plt.imshow(unsharp mask)
plt.title('Unsharp Masking')
plt.axis('off')
plt.subplot(1, 3, 3)
plt.imshow(high boost)
plt.title('High Boost Filtering (k=3)')
plt.axis('off')
```

plt.show()







Q2. Sharpen the following image using the concept of Laplacian Filtering.

CODE

```
img = cv2.imread('/kaggle/input/lab-05-dataset/lab 05/Picture2.jpg')
img gray = cv2.cvtColor(img, cv2.COLOR BGR2GRAY)
laplacian = cv2.Laplacian(img_gray, ddepth=cv2.CV_64F, ksize=3)
laplacian = np.uint8(np.absolute(laplacian))
sharpened = cv2.add(img gray, laplacian)
plt.figure(figsize=(15, 5))
plt.subplot(1, 3, 1)
plt.imshow(img_gray, cmap='gray')
plt.title('Original (Grayscale)')
plt.axis('off')
plt.subplot(1, 3, 2)
plt.imshow(laplacian, cmap='gray')
plt.title('Laplacian Filtered')
plt.axis('off')
plt.subplot(1, 3, 3)
plt.imshow(sharpened, cmap='gray')
plt.title('Sharpened Image (Laplacian)')
plt.axis('off')
plt.tight layout()
plt.show()
```

Original (Grayscale)



Laplacian Filtered



Sharpened Image (Laplacian)

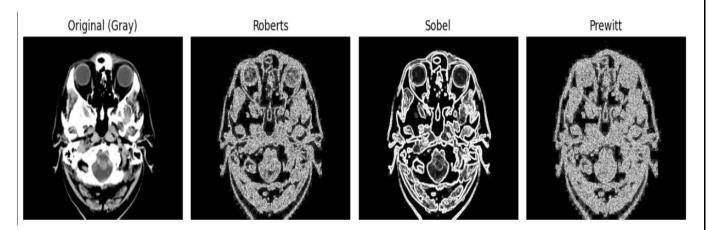


Q3. Use Roberts-cross, Sobel, and Prewitt operators to detect the edge of the following image.

CODE

```
from scipy.ndimage import convolve
img = cv2.imread('/kaggle/input/lab-05-dataset/lab 05/Picture3.jpg')
gray = cv2.cvtColor(img, cv2.COLOR BGR2GRAY)
roberts cross v = np.array([[1, 0],
                 [0, -1], dtype=int)
roberts cross h = np.array([[0, 1],
                 [-1, 0]], dtype=int)
roberts v = convolve(gray, roberts cross v)
roberts h = convolve(gray, roberts cross h)
roberts = np.sqrt(roberts v^{**}2 + roberts h^{**}2)
roberts = np.uint8(np.clip(roberts, 0, 255))
sobelx = cv2.Sobel(gray, cv2.CV 64F, 1, 0, ksize=3)
sobely = cv2.Sobel(gray, cv2.CV 64F, 0, 1, ksize=3)
sobel = np.sqrt(sobelx**2 + sobely**2)
sobel = np.uint8(np.clip(sobel, 0, 255))
prewitt kernelx = np.array([[1, 0, -1],
                 [1, 0, -1],
                 [1, 0, -1]], dtype=int)
prewitt kernely = np.array([[1, 1, 1],
                 [0, 0, 0],
                 [-1, -1, -1]], dtype=int)
prewittx = convolve(gray, prewitt kernelx)
prewitty = convolve(gray, prewitt kernely)
prewitt = np.sqrt(prewittx**2 + prewitty**2)
prewitt = np.uint8(np.clip(prewitt, 0, 255))
titles = ['Original (Gray)', 'Roberts', 'Sobel', 'Prewitt']
images = [gray, roberts, sobel, prewitt]
plt.figure(figsize=(12, 6))
for i in range(4):
  plt.subplot(1, 4, i + 1)
  plt.imshow(images[i], cmap='gray')
  plt.title(titles[i])
```

plt.axis('off')
plt.tight_layout()
plt.show()



Q4. Show performance comparison among High Boost, Unsharp, Laplacian Roberts-cross, Sobel,

Prewitt and Canny filtering for edge detection – find out which one is better for the given image.

CODE

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
import os
def apply edge detectors(image path):
  img_name = os.path.splitext(os.path.basename(image_path))[0] #
e.g., 'Picture1'
  img = cv2.imread(image path, cv2.IMREAD GRAYSCALE)
  img = cv2.GaussianBlur(img, (3, 3), 0)
  # Sharpening
  blurred = cv2.GaussianBlur(img, (0, 0), 3)
  unsharp = cv2.addWeighted(img, 1.5, blurred, -0.5, 0)
  high boost = cv2.addWeighted(img, 2.5, blurred, -1.5, 0)
  # Laplacian
  laplacian = cv2.Laplacian(img, cv2.CV 64F)
  laplacian = np.uint8(np.absolute(laplacian))
```

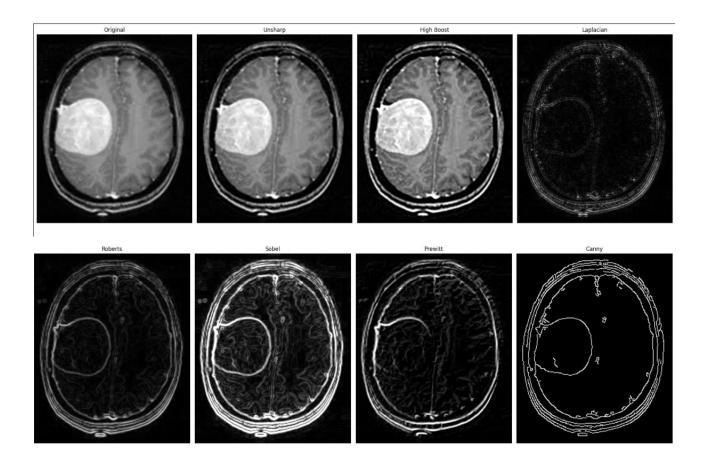
Roberts

kx = np.array([[1, 0], [0, -1]], dtype=np.float32)ky = np.array([[0, 1], [-1, 0]], dtype=np.float32)

```
robertsx = cv2.filter2D(img, cv2.CV 64F, kx)
  robertsy = cv2.filter2D(img, cv2.CV 64F, ky)
  roberts = np.sqrt(robertsx**2 + robertsy**2)
  roberts = np.uint8(np.clip(roberts, 0, 255))
  # Sobel
  sobelx = cv2.Sobel(img, cv2.CV 64F, 1, 0, ksize=3)
  sobely = cv2.Sobel(img, cv2.CV 64F, 0, 1, ksize=3)
  sobel = np.sqrt(sobelx**2 + sobely**2)
  sobel = np.uint8(np.clip(sobel, 0, 255))
  # Prewitt
  prewittx = cv2.filter2D(img, -1, np.array([[-1, 0, 1], [-1, 0, 1], [-1, 0, 1], [-1, 0, 1])
1]]))
  prewitty = cv2.filter2D(img, -1, np.array([[-1, -1, -1], [0, 0, 0], [1, 1, -1]))
1]]))
  prewitt = np.sqrt(prewittx.astype(float)**2 +
prewitty.astype(float)**2)
  prewitt = np.uint8(np.clip(prewitt, 0, 255))
  # Canny
  canny = cv2.Canny(img, 100, 200)
  # Save to /kaggle/working/
  save dir = "/kaggle/working/"
  results = {
     "unsharp": unsharp,
     "high boost": high boost,
     "laplacian": laplacian,
     "roberts": roberts,
    "sobel": sobel,
     "prewitt": prewitt,
     "canny": canny
  for key, image in results.items():
    save path = os.path.join(save dir, f"{img name} {key}.jpg")
     cv2.imwrite(save path, image)
  plt.figure(figsize=(20, 15))
  titles = ["Original", "Unsharp", "High Boost", "Laplacian", "Roberts",
"Sobel", "Prewitt", "Canny"]
  images = [img, unsharp, high boost, laplacian, roberts, sobel, prewitt,
canny]
  for i in range(8):
```

```
plt.subplot(2, 4, i + 1)
  plt.imshow(images[i], cmap='gray')
  plt.title(titles[i])
  plt.axis('off')
plt.tight_layout()
plt.show()
```

apply_edge_detectors("/kaggle/input/lab-05-dataset/lab_05/Picture1.jpg")
apply_edge_detectors("/kaggle/input/lab-05-dataset/lab_05/Picture2.jpg")
apply_edge_detectors("/kaggle/input/lab-05-dataset/lab_05/Picture3.jpg")









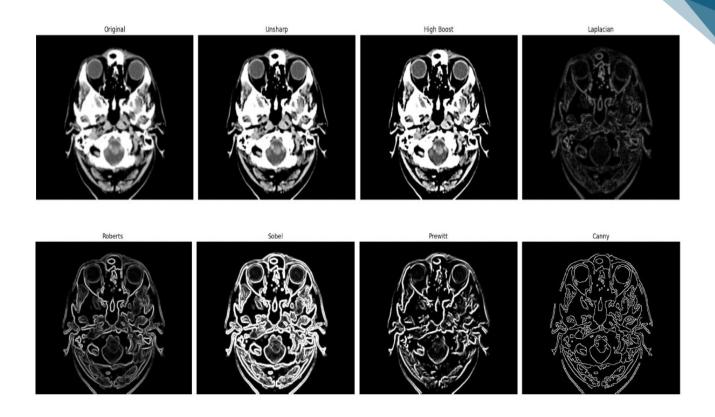












-----Comparison Table-----

Filter	Strengths	Weaknesses	Best For
Unsharp Masking	Enhances edges subtly	Weak on noisy images	Mild sharpening
High Boost	Stronger edge enhancement	Amplifies noise	High-contrast edges
Laplacian	Detects fine edges	Very noise-sensitive	Sharp transitions
Roberts Cross	Fast, simple	Poor noise handling	Binary edges / small images
Sobel	Smooth gradients	Thick edges	General-purpose detection
Prewitt	Similar to Sobel	Less noise-resistant	Basic edge detection
Canny	Clean, thin edges	Slower computation	High-quality edge maps

In a short Summary:

For sharpening:

- → Unsharp Masking for subtle,
- → **High Boost** for strong enhancement.

For edge detection:

- \rightarrow Canny gives the cleanest, most accurate edges.
- \rightarrow **Sobel** is a good faster alternative.

Best Overall:

Canny for edge detection, High Boost for sharpening