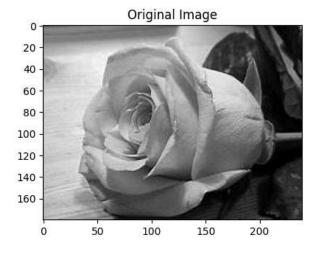
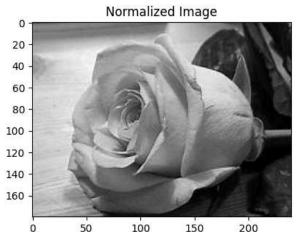
# Lab<sub>1</sub>

#### 1. Image Preprocessing

- **Algorithm**: Normalized a grayscale image to the range [0, 1].
- Accuracy: Ensures that pixel values are properly scaled for further processing.
- Visualization: Displayed the original and normalized images.

```
In [5]: import numpy as np
        import cv2
        import matplotlib.pyplot as plt
        # Load a grayscale image
        image = cv2.imread('rose.jpeg', cv2.IMREAD_GRAYSCALE)
        # Normalize the image to the range [0, 1]
        normalized_image = image / 255.0
        # Visualization
        plt.figure(figsize=(10, 5))
        plt.subplot(1, 2, 1)
        plt.title('Original Image')
        plt.imshow(image, cmap='gray')
        plt.subplot(1, 2, 2)
        plt.title('Normalized Image')
        plt.imshow(normalized_image, cmap='gray')
        plt.show()
```

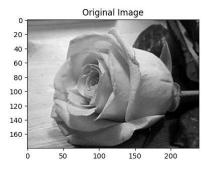


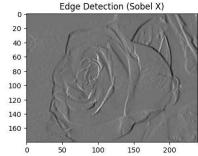


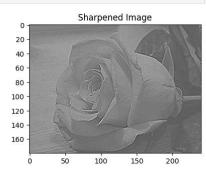
## 2. Apply Convolution

- Algorithm: Applied Sobel filter (X direction) for edge detection and a sharpening filter.
- **Accuracy**: Use filters to emphasize certain features.
- **Visualization**: Shown the original, edge-detected, and sharpened images.

```
In [6]: # Define Sobel X kernel and Sharpening kernel
        sobel x = np.array([[-1, 0, 1],
                            [-2, 0, 2],
                            [-1, 0, 1]
        sharpen_kernel = np.array([[0, -1, 0],
                                    [-1, 5, -1],
                                    [0, -1, 0]
        # Apply convolution using Sobel filter (X direction)
        edge x = cv2.filter2D(normalized image, -1, sobel x)
        # Apply convolution using Sharpening filter
        sharpened image = cv2.filter2D(normalized image, -1, sharpen kernel)
        # Visualization
        plt.figure(figsize=(15, 5))
        plt.subplot(1, 3, 1)
        plt.title('Original Image')
        plt.imshow(normalized image, cmap='gray')
        plt.subplot(1, 3, 2)
        plt.title('Edge Detection (Sobel X)')
        plt.imshow(edge x, cmap='gray')
        plt.subplot(1, 3, 3)
        plt.title('Sharpened Image')
        plt.imshow(sharpened image, cmap='gray')
        plt.show()
```







# 3. Apply Correlation

- **Algorithm**: Applied correlation to the grayscale and convolved images.
- **Accuracy**: Measure the similarity between the original and filtered images.
- **Visualization**: Not applicable, as this is a numerical operation.

```
In [7]: # Function to compute correlation
def compute_correlation(image1, image2):
    return np.corrcoef(image1.flatten(), image2.flatten())[0, 1]

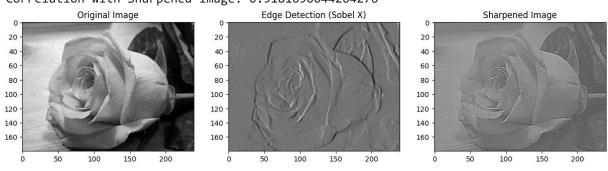
# Correlation between original and convolved images
corr_edge = compute_correlation(normalized_image, edge_x)
corr_sharpen = compute_correlation(normalized_image, sharpened_image)

print(f'Correlation with Edge Detected Image: {corr_edge}')
```

```
print(f'Correlation with Sharpened Image: {corr_sharpen}')

# Visualization
plt.figure(figsize=(15, 5))
plt.subplot(1, 3, 1)
plt.title('Original Image')
plt.imshow(normalized_image, cmap='gray')
plt.subplot(1, 3, 2)
plt.title('Edge Detection (Sobel X)')
plt.imshow(edge_x, cmap='gray')
plt.subplot(1, 3, 3)
plt.title('Sharpened Image')
plt.imshow(sharpened_image, cmap='gray')
plt.show()
```

Correlation with Edge Detected Image: 0.009757842549018566 Correlation with Sharpened Image: 0.9181090644264276



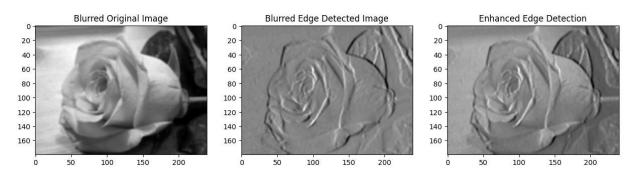
## 4. Edge Detection and Enhancement

- **Algorithm**: Applied Gaussian blur to the original and edge-detected images to enhance edges.
- **Accuracy**: The blur can help smooth noise while enhancing important features.
- Visualization: Shown the blurred images and the enhanced edge-detected image.

```
In [8]: # Apply Gaussian Blur to original and edge-detected images
        blurred_original = cv2.GaussianBlur(normalized_image, (5, 5), 0)
        blurred_edge = cv2.GaussianBlur(edge_x, (5, 5), 0)
        # Enhanced edge detection (combining original blurred with edge-detected blurred)
        enhanced_edge = cv2.addWeighted(blurred_original, 0.5, blurred_edge, 0.5, 0)
        # Visualization
        plt.figure(figsize=(15, 5))
        plt.subplot(1, 3, 1)
        plt.title('Blurred Original Image')
        plt.imshow(blurred_original, cmap='gray')
        plt.subplot(1, 3, 2)
        plt.title('Blurred Edge Detected Image')
        plt.imshow(blurred edge, cmap='gray')
        plt.subplot(1, 3, 3)
        plt.title('Enhanced Edge Detection')
        plt.imshow(enhanced edge, cmap='gray')
        plt.show()
```

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### 5. Comparison and Analysis

- Interpretation:
  - Correlation Results: Higher correlation indicates greater similarity between images.
  - **Effect of Gaussian Blur**: Gaussian blur helps reduce noise while enhancing critical features when combined with edge detection, leading to a cleaner and more pronounced edge detection result.

In [9]: print(f"Correlation between the original and edge-detected images: {corr\_edge:.4f}"
 print(f"Correlation between the original and sharpened images: {corr\_sharpen:.4f}")
 print("The effect of combining edge detection with Gaussian blur results in a clean

Correlation between the original and edge-detected images: 0.0098 Correlation between the original and sharpened images: 0.9181 The effect of combining edge detection with Gaussian blur results in a cleaner, enhanced edge-detected image, reducing noise and emphasizing important features.