cat1

April 8, 2025

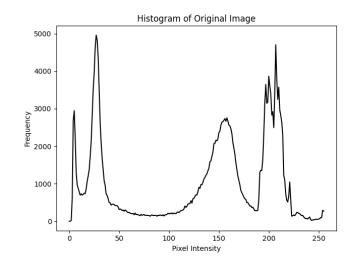
QUESTION 1

plt.tight_layout()

plt.show()

```
[1]: %%capture --no-stderr
     %pip install --quiet numpy matplotlib scikit-image opencv-python
[]: \#(i)Load a sample grayscale image in a Jupyter Notebook and analyze its
     histogram. Display both the image and its histogram side by side.
     import numpy as np
     import matplotlib.pyplot as plt
     from skimage import data, exposure
     # Load a sample grayscale image (the "camera" image)
     image = data.camera()
     # Compute the histogram
     hist, bins = np.histogram(image.flatten(), bins=256, range=[0,256])
     # Display image and histogram side by side
     fig, axes = plt.subplots(1, 2, figsize=(12, 5))
     axes[0].imshow(image, cmap='gray')
     axes[0].set_title('Original Grayscale Image')
     axes[0].axis('off')
     axes[1].plot(bins[:-1], hist, color='black')
     axes[1].set_title('Histogram of Original Image')
     axes[1].set_xlabel('Pixel Intensity')
     axes[1].set_ylabel('Frequency')
```





```
[2]: # (ii) Modify the contrast of the image (increase and decrease it) and generate,
      → the histograms for each version. Compare the histograms and explain your
      ⇔observations.
     # Increase contrast using contrast stretching
     p2, p98 = np.percentile(image, (2, 98))
     image high contrast = exposure.rescale intensity(image, in range=(p2, p98))
     # Decrease contrast by compressing the intensity range e.g. by mapping the
      ⇔intensities to a smaller range
     image_low_contrast = exposure.rescale_intensity(image, in_range=(0, 255),__
     →out_range=(100, 150))
     # Compute histograms for the modified images
     hist_high, bins_high = np.histogram(image_high_contrast.flatten(), bins=256,__
      →range=[0,256])
     hist_low, bins_low = np.histogram(image_low_contrast.flatten(), bins=256,__
      →range=[0,256])
     # Display images and histograms
     fig, axes = plt.subplots(2, 2, figsize=(12, 10))
     # Original image and histogram for reference
     axes[0,0].imshow(image, cmap='gray')
     axes[0,0].set_title('Original Image')
     axes[0,0].axis('off')
     axes[0,1].plot(bins[:-1], hist, color='black')
     axes[0,1].set_title('Histogram of Original Image')
     axes[0,1].set_xlabel('Intensity')
     axes[0,1].set_ylabel('Frequency')
```

```
# High contrast image and its histogram
axes[1,0].imshow(image_high_contrast, cmap='gray')
axes[1,0].set_title('High Contrast Image')
axes[1,0].axis('off')
axes[1,1].plot(bins_high[:-1], hist_high, color='blue')
axes[1,1].set_title('Histogram of High Contrast Image')
axes[1,1].set_xlabel('Intensity')
axes[1,1].set_ylabel('Frequency')
plt.tight_layout()
plt.show()
# For the low contrast image, we can plot separately:
fig, axes = plt.subplots(1, 2, figsize=(12, 5))
axes[0].imshow(image_low_contrast, cmap='gray')
axes[0].set_title('Low Contrast Image')
axes[0].axis('off')
axes[1].plot(bins_low[:-1], hist_low, color='red')
axes[1].set_title('Histogram of Low Contrast Image')
axes[1].set_xlabel('Intensity')
axes[1].set_ylabel('Frequency')
plt.tight_layout()
plt.show()
```



Histogram of Original Image

5000

4000

1000

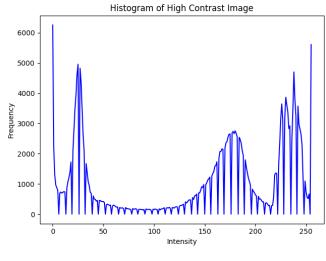
500

150

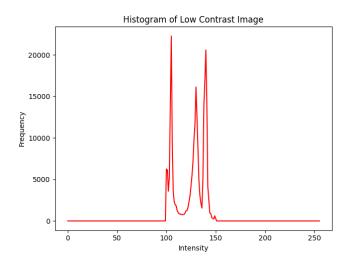
2000

250

High Contrast Image







OBSERVATIONS

High Contrast: The histogram of the high contrast image tends to be more "spread out" across the intensity range (i.e. a wider spread of pixel values) compared to the original one.

Low Contrast: The low contrast image's histogram is compressed into a narrow band (in our example, between intensity values 100 and 150). The pixel values are concentrated over a smaller range, making the image appear "washed out" or flat.

Question 3

(i)

Morphological operations are non-linear transformations applied to binary or grayscale images that process structures based on their shape. These operations use a probe called a structuring element (SE) to inspect and modify the geometrical structure of an image. Useful in shape analysis, noise removal, image segmentation, and object extraction.

```
[4]: # a) Dilation

# Effect: Expands white regions (foreground), fills small holes, bridges gaps.

dilated = cv2.dilate(img, kernel)

# b) Erosion

# Effect: Shrinks white regions (foreground), removes small objects or noise.

eroded = cv2.erode(img, kernel)

# c) Opening

# Effect: Smoothes contours, removes small foreground noise.

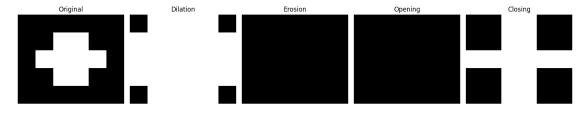
opening = cv2.morphologyEx(img, cv2.MORPH_OPEN, kernel)
```

```
# d) Closing
# Effect: Closes small holes, connects narrow breaks.
closing = cv2.morphologyEx(img, cv2.MORPH_CLOSE, kernel)
```

```
[5]: #Visualisation

titles = ['Original', 'Dilation', 'Erosion', 'Opening', 'Closing']
images = [img, dilated, eroded, opening, closing]

plt.figure(figsize=(15, 3))
for i in range(5):
    plt.subplot(1, 5, i + 1)
    plt.imshow(images[i], cmap='gray')
    plt.title(titles[i])
    plt.axis('off')
plt.tight_layout()
plt.show()
```



```
plt.figure(figsize=(15, 3))
for i in range(5):
    plt.subplot(1, 5, i + 1)
    plt.imshow(images[i], cmap='gray')
    plt.title(titles[i])
    plt.axis('off')
plt.tight_layout()
plt.show()
```

