Lab Report-03

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Submitted to-

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Lab Report Title

Image Enhancement Using Histogram Equalization in Python

Introduction

Histogram equalization is a method for enhancing the contrast of an image by spreading out the intensity values across the entire range. This technique increases the global contrast of images, especially when their data is concentrated in a narrow range, improving visibility and highlighting details. Histogram equalization is widely used in fields such as medical imaging, remote sensing, and computer vision.

Python code

```
# histogram_equalization.py
from PIL import Image
import numpy as np
import matplotlib.pyplot as plt
def histogram_equalization(input_path, output_path):
    # Load the image and convert to grayscale
    img = Image.open(input_path).convert("L")
    img_array = np.array(img)
    # Flatten the image array and calculate histogram
    hist, bins = np.histogram(img_array.flatten(), 256, [0,
       \hookrightarrow 256])
    # Calculate cumulative distribution function (CDF)
    cdf = hist.cumsum()
    cdf_normalized = cdf * 255 / cdf[-1] # Normalize the
       \hookrightarrow CDF to [0, 255]
    # Apply histogram equalization
    img_equalized = np.interp(img_array.flatten(),
       → bins[:-1], cdf_normalized)
    img_equalized =

    img_equalized.reshape(img_array.shape).astype(np.uint8)

    equalized_img = Image.fromarray(img_equalized)
    # Save and display the images
    equalized_img.save(output_path)
    # Plot original and equalized images with histograms
    plt.figure(figsize=(12, 6))
```

```
# Original Image and Histogram
    plt.subplot(2, 2, 1)
    plt.imshow(img, cmap="gray")
    plt.title("Original Image")
    plt.axis("off")
    plt.subplot(2, 2, 2)
    plt.hist(img_array.flatten(), 256, [0, 256],

    color='black')

    plt.title("Histogram of Original Image")
    # Equalized Image and Histogram
    plt.subplot(2, 2, 3)
    plt.imshow(equalized_img, cmap="gray")
    plt.title("Equalized Image")
    plt.axis("off")
    plt.subplot(2, 2, 4)
    plt.hist(img_equalized.flatten(), 256, [0, 256],

    color='black')

    plt.title("Histogram of Equalized Image")
    plt.tight_layout()
    plt.show()
# Example usage
if __name__ == "__main__":
    input_path = "input.jpg"
                                       # Replace with the
       \hookrightarrow path to your input image
    output_path = "equalized.jpg"  # Path for saving the
       \hookrightarrow equalized image
    histogram_equalization(input_path, output_path)
    print(f"Equalized image saved as {output_path}")
```

Input



Figure 1: Input Image

Output





Equalized image saved as equalized.jpg

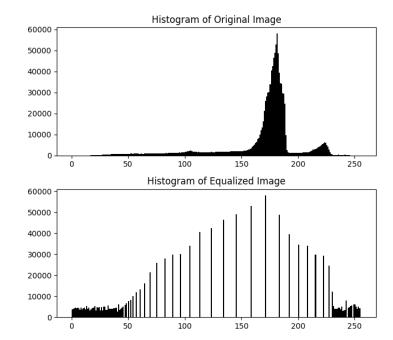


Figure 2: Output

Lab Report Title

Image Enhancement Using Histogram Specification in Python

Introduction

Histogram specification (or histogram matching) is a process of adjusting the pixel intensity distribution of an image to match a specified histogram. Unlike histogram equalization, which enhances contrast uniformly, histogram specification allows more control by transforming an image's histogram to resemble that of a reference image. This technique is often used in applications where visual consistency across different images is important, such as medical imaging or remote sensing.

Python code

```
# histogram_specification.py
from PIL import Image
import numpy as np
import matplotlib.pyplot as plt
def match_histograms(source_img, reference_img):
    # Convert images to arrays
    source_array = np.array(source_img)
    reference_array = np.array(reference_img)
    # Flatten the image arrays and calculate histograms
    source_hist, bin_edges =
       → np.histogram(source_array.flatten(), bins=256,
       \rightarrow range=(0, 256))
    reference_hist, _ =
       → np.histogram(reference_array.flatten(), bins=256,
       \hookrightarrow range=(0, 256))
    # Calculate cumulative distribution functions (CDFs)
    source_cdf = source_hist.cumsum()
    source_cdf = source_cdf / source_cdf[-1]
    reference_cdf = reference_hist.cumsum()
    reference_cdf = reference_cdf / reference_cdf[-1]
    # Create a lookup table to match source CDF to reference
    lookup_table = np.interp(source_cdf, reference_cdf,
       \hookrightarrow bin_edges[:-1])
    # Map the source image pixels using the lookup table
```

```
matched_array = np.interp(source_array.flatten(),
      \hookrightarrow bin_edges[:-1],
      → lookup_table).reshape(source_array.shape)
    matched_img =
      → Image.fromarray(matched_array.astype(np.uint8))
    return matched_img
# Display and save images
def display_and_save_images(source_path, reference_path,
  \hookrightarrow output_path):
    # Load images
    source_img = Image.open(source_path).convert("L")
    reference_img = Image.open(reference_path).convert("L")
    # Perform histogram specification
    matched_img = match_histograms(source_img, reference_img)
    # Save the matched image
    matched_img.save(output_path)
    # Display the images
    plt.figure(figsize=(15, 5))
    # Original source image
    plt.subplot(1, 3, 1)
    plt.imshow(source_img, cmap="gray")
    plt.title("Source Image")
    plt.axis("off")
    # Reference image
    plt.subplot(1, 3, 2)
    plt.imshow(reference_img, cmap="gray")
    plt.title("Reference Image")
    plt.axis("off")
    # Matched image
    plt.subplot(1, 3, 3)
    plt.imshow(matched_img, cmap="gray")
    plt.title("Matched Image")
    plt.axis("off")
    plt.tight_layout()
    plt.show()
# Example usage
if __name__ == "__main__":
```

Input



Figure 3:

Output



Figure 4: