Task: Implementing Thresholding

Student Information

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1. Fixed Thresholding Fixed thresholding is a simple method where we choose a threshold value, and any pixel intensity value above the threshold is set to one value (e.g., white), and any pixel intensity below the threshold is set to another value (e.g., black).

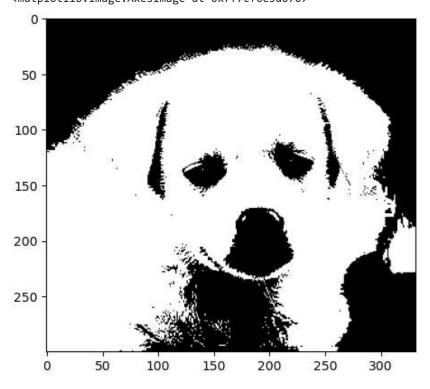
```
import cv2
import numpy as np
import matplotlib.pyplot as plt

image = cv2.imread('./Dog.jpg', cv2.IMREAD_GRAYSCALE)

# Apply Fixed Thresholding
_, fixed_thresh = cv2.threshold(image, 127, 255, cv2.THRESH_BINARY)

plt.imshow(fixed_thresh, cmap='gray')
```

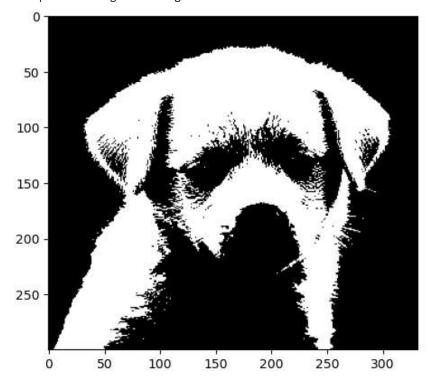
<matplotlib.image.AxesImage at 0x7f7cf6e5d070>



2. OTSU's Thresholding OTSU's method automatically calculates the optimal threshold value for a bimodal image (an image with two peaks in its histogram).

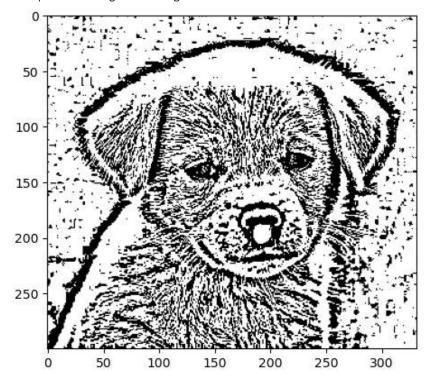
```
# Apply OTSU's Thresholding
_, otsu_thresh = cv2.threshold(image, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
plt.imshow(otsu_thresh, cmap='gray')
```

<matplotlib.image.AxesImage at 0x7f7d2d11f0b0>



3. Adaptive Mean Thresholding In Adaptive Mean Thresholding, the threshold value is the mean of the neighborhood area minus a constant.

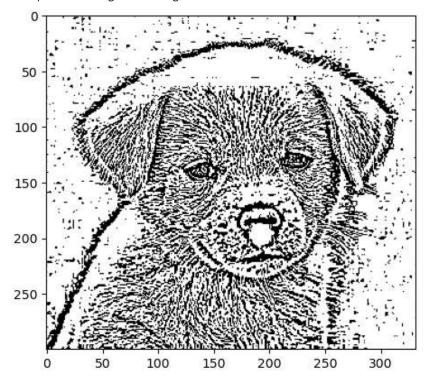
<matplotlib.image.AxesImage at 0x7f7d2c0efda0>



4. Adaptive Gaussian Thresholding In Adaptive Gaussian Thresholding, the threshold value is a weighted sum (Gaussian window) of the neighborhood values minus a constant.

```
plt.imshow(adaptive_gaussian_thresh, cmap='gray')
```

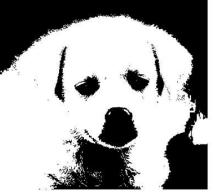
<matplotlib.image.AxesImage at 0x7f7cf6e5d190>



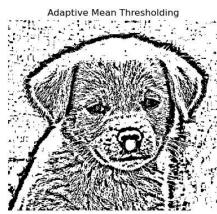
```
# Plotting the results
plt.figure(figsize=(12, 8))
plt.subplot(2, 3, 1)
plt.imshow(image, cmap='gray')
plt.title('Original Image')
plt.axis('off')
plt.subplot(2, 3, 2)
plt.imshow(fixed_thresh, cmap='gray')
plt.title('Fixed Thresholding')
plt.axis('off')
plt.subplot(2, 3, 3)
plt.imshow(otsu_thresh, cmap='gray')
plt.title("OTSU's Thresholding")
plt.axis('off')
plt.subplot(2, 3, 4)
plt.imshow(adaptive_mean_thresh, cmap='gray')
plt.title('Adaptive Mean Thresholding')
plt.axis('off')
plt.subplot(2, 3, 5)
plt.imshow(adaptive_gaussian_thresh, cmap='gray')
plt.title('Adaptive Gaussian Thresholding')
plt.axis('off')
plt.tight_layout()
plt.show()
```

Original Image Fixed Thresholding OTSU's Thresholding











Summary:

- Fixed Thresholding is straightforward but requires a pre-defined threshold value.
- OTSU's Thresholding is optimal for bimodal images.
- Adaptive Mean and Adaptive Gaussian Thresholding are better for images with varying lighting conditions.

Task: Manually Implementing Thresholding Methods

1. Fixed Thresholding (Manual Implementation)

```
image = cv2.imread('./Dog.jpg', cv2.IMREAD_GRAYSCALE)

def fixed_threshold(image, threshold):
    return np.where(image > threshold, 255, 0).astype(np.uint8)

fixed_thresh = fixed_threshold(image, 127)
```

2. OTSU's Thresholding (Manual Implementation)

```
def otsu_threshold(image):
    pixel_counts = np.bincount(image.ravel(), minlength=256)
    s_max = (0,-10)
    for threshold in range(256):
        w0 = sum(pixel_counts[:threshold])
        w1 = sum(pixel_counts[threshold:])
        if w0 == 0 or w1 == 0:
            continue
        mu0 = sum([i * pixel_counts[i] for i in range(0, threshold)]) / w0
        mu1 = sum([i * pixel_counts[i] for i in range(threshold, 256)]) / w1
        # Inter-class variance
        s = w0 * w1 * (mu0 - mu1) ** 2
```

3. Adaptive Mean Thresholding (Manual Implementation)

```
def adaptive_mean_threshold(image, block_size, C):
    half_block = block_size // 2
    padded_image = np.pad(image, pad_width=half_block, mode='constant', constant_values=0)
    adaptive_thresh = np.zeros_like(image)

for i in range(half_block, padded_image.shape[0] - half_block):
    for j in range(half_block, padded_image.shape[1] - half_block):
        local_block = padded_image[i-half_block:i+half_block+1, j-half_block:j+half_block+1]
        local_thresh = np.mean(local_block) - C
        adaptive_thresh[i-half_block, j-half_block] = 255 if image[i-half_block, j-half_block] > ...
    return adaptive_thresh

adaptive_mean_thresh = adaptive_mean_threshold(image, 11, 2)
```

4. Adaptive Gaussian Thresholding (Manual Implementation)

```
from scipy.ndimage import gaussian_filter

def adaptive_gaussian_threshold(image, block_size, C):
    half_block = block_size // 2
    padded_image = np.pad(image, pad_width=half_block, mode='constant', constant_values=0)
    adaptive_thresh = np.zeros_like(image)

gaussian_image = gaussian_filter(padded_image, sigma=block_size/6.0)

for i in range(half_block, padded_image.shape[0] - half_block):
    for j in range(half_block, padded_image.shape[1] - half_block):
        local_thresh = gaussian_image[i, j] - C
        adaptive_thresh[i-half_block, j-half_block] = 255 if image[i-half_block, j-half_block] >
    return adaptive_thresh

adaptive_gaussian_thresh = adaptive_gaussian_threshold(image, 11, 2)
```

Visualizing the Results

```
# Plotting the results
plt.figure(figsize=(12, 8))

plt.subplot(2, 3, 1)
plt.imshow(image, cmap='gray')
plt.title('Original Image')
plt.axis('off')

plt.subplot(2, 3, 2)
plt.imshow(fixed_thresh, cmap='gray')
plt.title('Fixed Thresholding')
plt.axis('off')

plt.subplot(2, 3, 3)
plt.imshow(otsu_thresh, cmap='gray')
plt.title("OTSU's Thresholding")
plt.axis('off')

plt.subplot(2, 3, 4)
```

```
plt.imshow(adaptive_mean_thresh, cmap='gray')
plt.title('Adaptive Mean Thresholding')
plt.axis('off')

plt.subplot(2, 3, 5)
plt.imshow(adaptive_gaussian_thresh, cmap='gray')
plt.title('Adaptive Gaussian Thresholding')
plt.axis('off')

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







Adaptive Mean Thresholding





Explanation:

- Fixed Thresholding is implemented by setting a pixel value to 255 if it's above a certain threshold, and 0 otherwise.
- OTSU's Thresholding is manually computed by finding the optimal threshold that maximizes the inter-class variance
- Adaptive Mean Thresholding calculates the mean of a local block around each pixel and subtracts a constant to set the threshold.
- Adaptive Gaussian Thresholding uses a Gaussian-weighted sum for the local block to determine the threshold.